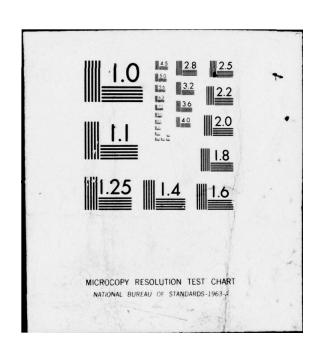
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TECHNICAL REPORT 76-10

SELECTING ANALYTIC APPROACHES FOR DECISION SITUATIONS

A MATCHING OF TAXONOMIES

by

Rex V. Brown Jacob W. Ulvila

Sponsored by

Office of Naval Research Engineering Psychology Programs Contract No. N00014-75-C-0426

October, 1976



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and practitioners identify appropriate analytic approaches for any given decision situation. An attempt is made to suggest a taxonomic framework for codifying the state-of-the-art of decision analysis, a language for expressing matching generalizations which associate the appropriate analytic option to a particular situation. This language has three main components, each of which has an exhaustive numerical coding scheme.

The first component is a "situation taxonomy," listing about one hundred dimensions of a situation that might be relevant to a particular analytic choice. These dimensions include: the stakes involved in a decision; the reaction time available; and the clarity with which options, probable consequences, and values are perceived.

The second component is an "analysis taxonomy," according to which about one hundred decision-analytic choices can be located in an "analytic option space." Dimensions of the analytic taxonomy include: how much decision analysis is undertaken, how it is used, what type of model structure is involved, and what technique for probability assessment or consequence evaluation is employed.

The third component is a "performance measure taxonomy," listing about thirty measures of effectiveness which can characterize the analytic options. The same taxonomy can also be used to describe a situation by expressing the relative importance of the performance measures in the situation. Performance measure dimensions include: enhanced logical reasoning, cost, speed, convenience, and facilitated communication. This component serves as a mediating factor, implicit or explicit, in matching analysis to situation.

In this research effort, we have attempted to identify a few important and plausible matching generalizations based on the experience of practicing decision analysts. A few analytic options were selected to represent thousands of possibilities and to facilitate generalizations about when they should be exercised in the form of a taxonomy matching. A U.S. decision on whether to export high-technology items to the Soviet Bloc is analyzed by using the taxonomic matching framework. Other illustrative material is also used throughout the report.

SUMMARY

This report attempts to develop a conceptual framework within which the state-of-the-art of applied decision analysis can be codified. The framework consists of a three-way taxonomy: decision situations, analytic options, and performance measures. Within this framework, a tentative and illustrative set of practical guidelines is presented to help decision analysis users and practioners identify appropriate analytic approaches for any given decision situation.

Decision makers and decision analysis practitioners are both faced with problems of determining what type of analysis, if any, to apply to a given decision situation. The standard approach in such cases has been for the decision maker or analyst to evaluate, case by case, any special features of the situation and for him to apply informal experience and judgment, reflecting his perception of the state-of-the-art of decision analysis, to determine the appropriate approach for the situation.

In this report, an attempt is made to suggest a taxonomic framework for codifying the state-of-the-art of decision analysis, a language for expressing "matching generalizations" which associate the appropriate analytic option to a particular situation. This language has three main components, each of which has an exhaustive numerical coding scheme.

The first component is a "situation taxonomy," listing about one hundred dimensions of a situation that might be relevant to a particular analytic choice. These dimensions include: the stakes involved in a decision; the reaction time available; and the clarity with which options, probable consequences, and values are perceived. These dimensions collectively define a "situation space" in which a situation can be unambiguously located.

The second component is an "analysis taxonomy," according to which about one hundred decision-analytic choices can be located in an "analytic option space." Dimensions of the analytic taxonomy include: how much decision analysis is undertaken, how it is used, what type of model structure is involved, and what technique for probability assessment or consequence evaluation is employed.

The third component is a "performance measure taxonomy," listing about thirty measures of effectiveness which can characterize the analytic options. The same taxonomy can

also be used to describe a situation by expressing the relative importance of the performance measures in the situation. Performance measure dimensions include: enhanced logical reasoning, cost, speed, convenience, and facilitated communication. This component serves as a mediating factor, implicit or explicit, in matching analysis to situation.

In this research effort, we have attempted to identify a few important and plausible matching generalizations based on the experience of practicing decision analysts. The generalizations do not attempt to be exhaustive, much less definitive. To do so would require formally codifying the entire current state-of-the-art of decision analysis. Instead, an attempt has been made to select a few analytic options to represent thousands of possibilities and to generalize about when they should be exercised in the form of a taxonomy matching. (They reflect the kind of technical judgment that an experienced decision analyst might communicate to a new decision analyst).

Two examples (representing informally the coded form of this report) are:

Generally consider using decision analysis on a single, non-recurring problem only if the "stakes" are more than a hundred thousand dollars, if the options are clearly visualized, and if the decision maker has more than usual difficulty in deciding among them.

Use the Delphi method for eliciting group probabilities only if there is a high degree of disparity in the status of group members and if the sources of information available to them are rather similar.

A U.S. decision on whether to export high-technology items to the Soviet Bloc is analyzed by using the taxonomic matching framework. Other illustrative material is also used throughout the report.

In addition to direct use of this "universal" situation taxonomy, the notion of "specific" situation taxonomies, which refer to specific fields such as naval tactics, is introduced. Each specific taxonomy can be more or less loosely associated with the universal situation taxonomy and hence matched to analytic options. The use of a specific taxonomy is illustrated in a characterization of government-versus-business decision situations.

At the current state of development, the situation taxonomy is too complex to be applied routinely in unfolding decision situations. The idea of using a procedure for progressively applying the taxonomy so that it will operate like a succession of progressively finer screens is presented.

Further developments suggested include: developing "filtering" decision rules (possibly for implementation on an interactive computer) that permit sequential (and economical) elicitation of situation characteristics, refining the taxonomy matching for selected analytic choices, and developing parochial taxonomies and matchings for selected fields (e.g., contingency planning for naval tactics).

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PREFACE

The present paper is a technical report on a study conducted over a two-year period, 1974-1976. The study represents a continuing task within the decision analysis research program sponsored by Dr. Martin A. Tolcott, Engineering Psychology Programs, Office of Naval Research.

The Decisions and Designs, Incorporated (DDI) project team consisted of Rex V. Brown, the principal investigator, and Jacob W. Ulvila, under the general direction of Cameron R. Peterson, DDI Technical Director. The report was edited by Michael L. Hays, DDI Publications Manager.

Since this project is visualized as a tentative first step in a major undertaking, at its grandest, a codification of the art of decision analysis, we would very much appreciate any comments or suggestions which may help produce something like a definitive framework for ourselves and others to work with. If responses are encouraging enough, this material may be adapted at some stage for general publication.

1.0 INTRODUCTION

Decision makers and analysts who are in a position to recommend the use of decision analysis are faced with problems of determining what type of analysis, if any, to apply to a given decision situation. Currently, the decision maker or analyst must rely upon his own experience and must make an implicit, case-by-case judgment of the appropriate analytic approach for each decision situation. This report is an initial attempt to improve this process by providing a codification of the combined judgments of experienced decision analysts.

1.1 Object and Background

Decision analysis might be defined as "a technology of making up your mind." Central to this technology is the argument that, given a set of reasonable behavior assumptions, any person's optimal decision can be unambiguously determined from quantified measures of his judgments, attitudes, and perceptions, by maximizing his subjective expected utility. This technology has enjoyed substantial vogue in the literature of management science over the past decade. As with most new technologies, however, the rate at which decision analysis has been applied in practice has been much slower than the rate at which its logical underpinnings have been developed. A reason for this lag is the lack of guidance that exists for identifying the most appropriate decision-analytic option to be used in a given situation.

The operational military commander or other decision maker has available to him, in principle, a moderately well developed array of decision-making aids based on decision analytic theory. However, it is by no means always readily apparent to him or, indeed, to his technical advisors when and where he can most advantageously use these tools and how. In theory, these tools can be applied to an almost indefinitely variable degree, to any situation involving choice or inference.

The broad purpose of this project, then, is to guide the decision maker and, to a lesser degree, his technical

See, for instance, Raiffa (1968), Schalifer (1969), and Brown, Kahr, and Peterson (1974); and Handbook for Decision Analysis (1973).

²See, for example, Section 4 of Brown <u>et al</u>. (1974).

staff, in determining the appropriate use of decisionanalytic tools. This paper is an attempt to produce some illustrative practical guidelines on the use of decision analysis. More importantly, it is an attempt to facilitate the advancement of the state-of-the-art of decision analysis by proposing a framework in which more definitive guidelines can be developed. It takes into account earlier, less extensive, attempts to classify or relate situations and analytic approaches.³

1.2 Terms of Reference

The main objective of this paper is to encourage and facilitate the codification of the current state-of-the-art of applied decision analysis. In order to meet this objective, this paper attempts to suggest a convenient taxonomy to classify a particular decision situation and a taxonomy of analytical options for addressing those situations, together with such guidance as can be given for an appropriate matching of the two.

Because an attempt to classify completely every conceivable decision-analytic option and every possible decision situation is an unmanageable task (with thousands of taxonomic characteristics for both analytic options and situations), the focus of this study is necessarily restricted. For now, we have chosen to restrict our main attention to those relatively few major dimensions which appear to be sufficient to characterize the important class of matching generalizations that are useful from the standpoint of a decision maker; only secondary consideration is given to developing guidelines for a decision analyst.

This restriction affects both the analytic taxonomy and the situation taxonomy. The scope of the analytic taxonomy is reduced because serious attention is devoted to the dozens of analytic options that are of interest to the decision maker rather than to the thousands of analytic options that are of interest to the analyst. The content of situation taxonomy is affected because the most appropriate characterization of a decision situation for specifying the correct options for the decision maker may differ from that appropriate for specifying the correct options for the analyst. For example, the fact that the decision situation calls for

³For example, by Howard (1968), Howard et al. (1975 and 1976), MacCrimmon and Taylor (1973), Payne et al. (1974), and Von Winterfeldt (1975).

formal documentation of the reasons behind a choice may be very relevant to deciding whether or not to use decision analysis at all (a decision maker's option) but almost irrelevant to whether simulation or backwards induction be used in the conduct of the decision analysis (an analyst's option).

Primary attention is also restricted to providing universal situation characterizations that allow matching rules to be developed for decision situations in general rather than for a particular class of decision situations, such as those faced by a businessman or those faced by a Navy task force This restriction affects the content of the commander. situation taxonomy. In particular, the situation taxonomy contains situation characterizations that facilitate matching analytic options and situations, rather than ones that make it easy for a decision maker to classify a situation. mately, to gain the maximum usefulness from the universal taxonomy, it will be necessary to develop taxonomies that enable situations to be more easily classified. is very large because decision makers in different fields, such as business or Navy tactics, tend to classify decision situations in different ways, with the result that a specialized taxonomy may be needed for each field. For example, a businessman may classify decision situations as investment decision situations, marketing decision situations, and so forth. Navy task force commander, on the other hand, may classify decisions as crisis decisions or wartime decisions. the specific taxonomy that will serve to relate the decision maker's preferred method of classifying situations to the method used in the universal taxonomy will be different for the businessman and for the task force commander. The development of the specific taxonomies, however, is beyond the scope of the present paper.

The present paper develops a set of matching generalizations in a universal taxonomy and illustrates these generalizations in the context of an actual decision situation, a determination of what level of embargo on the role of computers to the Soviet Bloc should be favored by the U.S. government. In addition, the concept of a specific taxonomy is notionally illustrated in the context of a Navy task force commander's decision.

1.3 Research Approach

In constructing a taxonomy of decision-making situations, it is natural to inquire whether principles exist for the construction of taxonomies in general. The most widespread use of taxonomic procedures has been in the biological sciences, and much has been written on appropriate ways to

produce classification systems.4 However, most of the stress in these writings is on the construction of automatic classification procedures. With these procedures, a dissimilarity measure is defined for pairs of objects that are to be classified, and then a clustering method is used to create classes. We do not feel that it is feasible to create such automatic classification procedures for decisionmaking situations, at least not at this early stage in our approach to the problem, since the similarity or dissimilarity between two decision-making situations is a matter of direct judgment. This similarity, at least at present, is unlikely to be derivable from a numerical formula. Other areas of study have used classification schemes, but there do not appear to be any principles in the literature for the construction of taxonomies that we can apply directly to decision-making situations.

The logic underlying the taxonomic matching of analytic options to situations is based upon the idea of maximizing the value of the decision analysis performed with due allowance for its cost. The formal principles of this idea, 5 are as yet cumbersome to apply in practice. Accordingly, the approach taken in this paper is rather informal, though implicitly it should be consistent with those principles.

The essential research approach used here was to examine the state-of-the-art as perceived by the authors and others and seek to codify the implicit taxonomic matching it represents. The illustrative methodological generalizations used were derived from the wide range of practical experience of analysts at Decision and Designs, Inc. In particular, the generalizations represent the judgments of two analysts (Rex V. Brown and Cameron R. Peterson), both of whom have actively applied decision analysis to a wide range of government and industry problems over the last eight years.

In addition, the situation taxonomy was checked and amplified by examining a large number of practical Naval and other decision situations. Experienced judgment was used to identify and prioritize their important distinguishing characteristics. Similarly, experienced judgment was applied to the classification of analytic options with a view toward making matching generalizations.

8

⁴See, for example, Sokal and Sneath (1963) or Jardine and Sibson (1971).

 $^{^{5}}$ Watson and Brown (1975/1).

2.0 A TAXONOMIC FRAMEWORK

In this study, an attempt is made to suggest a taxonomic framework that provides the terms for codifying the state-of-the-art of decision analysis. These terms provide a language for expressing "matching generalizations," which identify the appropriate analytic alternatives in any particular situation. This language has three main components.

The first component is a "situation taxonomy," which classifies a situation in terms relevant for choosing analytic alternatives. Representative dimensions of the situation taxonomy include: the stakes involved in a decision, the reaction time available, and the clarity with which options, probable consequences, and values are perceived. The complete set of these dimensions define a "situation space" in which a situation can be unambiguously located. The set of situation dimensions is summarized in Appendix A.

The second component is an "analysis taxonomy," which locates decision-analytic choices in an "analytic-option space." Representative dimensions of the analytic taxonomy include: how much decision analysis is undertaken; what role is assigned to it; what model structure is used; and what kind of probability assessment or consequence evaluation is applied. A summary of the dimensions of the analytic taxonomy is presented in Appendix B.

The third component is a "performance measure taxonomy," (Appendix C) which allows matching situations and analytic alternatives. The state-of-the-art of applied decision analysis can be thought of in terms of mapping one space onto another, that is, matching a situation with an analytic alternative (see Appendix F). The validation of such a mapping requires either implicit or explicit mediation of the performance measure taxonomy since its dimensions can be associated with situations as priorities and with analysis alternatives as properties. That is, although analytic options can always be matched to situations, the matchings cannot be justified or validated except by using the performance measure taxonomy. Representative dimensions of the performance measure taxonomy include: enhanced logical reasoning, lowered cost, and facilitated communication.

2.1 Situation Taxonomy

80

The situation taxonomy contains dimensions for classifying decision situations in a manner that facilitates identifying the appropriate amount of decision analysis and the appropriate decision-analytic choices for a given situation. This taxonomy includes only those situation distinctions, a

subset of all possible distinctions, impacting on those performance measures (see Section 2.3) that are important for determining the appropriate amount or type of decision analysis.

Table 2-1 contains a summary of questions (dimensions) in the situation taxonomy. Appendix A presents a detailed description of the taxonomic questions and possible responses (values) which between them define the situation characteristics.

Each situation characteristic is numerically coded for easy reference; thus, S112 refers to the question, "Expected Number of Occurrences," and S1123 refers to the answer, "Expected Number of Occurrences = 2" (defined in Appendix A). The "S" is used to distinguish the situation characteristics from other taxonomies. The final digit of the code is always a specific response to the question indicated by the other digits.

The items of Table 2-1 enclosed in parentheses are questions that distinguish only the type of decision analysis, not the appropriate amount of decision analysis, to be performed. Thus, these should be ignored in a first pass at characterizing a situation.

2.2 Analytic Options

The analytic taxonomy, summarized in Table 2-2 and detailed in Appendix B, attempts to identify many, but not all, decision-analytic choices available to decision makers and analysts and to classify them in a manner that facilitates a matching of specific alternatives to a decision situation. Future references to the analytic options will be preceded by the letter "A." Analogously to the situation taxonomy, the first digits signify the choice to be made; the last digit in the code corresponds to a specific alternative. For instance, "comprehensive/partial analysis?" will be referenced as A22 and "comprehensive analysis" as A221.

As with the situation taxonomy, any particular analysis will, in general, be characterized by a number of choices.

2.3 Performance Measures

The following list of performance measures attempts to capture the dimensions sufficient to determine the amount and type of decision analysis to use. To the extent that the relative importance of these measures can be determined directly in a given situation, the situation taxonomy presented

1 <u>De</u>	cision Substance	2	Decision Process
11 EZ	SIC SITUATION	21	REACTION TIME
(1)	1 current/contingent choice 2 expected # of occurrences 3 operating/information act) 9 other basic situation		211 minutes 212 hours 213 days 214 months
12 OF	TIONS	22	ANALYTIC PROCESSES
12	1 broad/narrow) 2 clear/fuzzy 3 complexity of decision options 4 radical/adaptive		221 # of input sources 222 analytic team 223 constraints on analytic method 224 documentation
13	5 static/dynamic 9 other options		229 other
13 DE	CIDABILITY	23	ORGANIZATIONAL PROCESSES
1 1:	difficulty of choice unfamiliarity key considerations do other decidability		(231 initiation) 232 responsibility (233 coordination) 234 justification 235 controversiality 236 performance control 237 rational-actor model
14 S	PAKES		238 risk attitude 239 other organizational processes
14	resources committed cost swing value swing maximum option impact expected irrationality cost other stakes	24	DECISION MAKER CHARACTERISTIC 241 role in organization 242 personal characteristics 249 other
.5 ot	TCOME VALUATION		
(1) (1) (1) (1)	<pre>3 measurable value?) 4 natural combinability of values) 5 timing)</pre>	25	RESOURCES AVAILABLE 251 computational facilities 252 staff 253 decision analysis expert 254 availability of decision maker (255 availability of assessors 256 dollars available 259 other resources
16 Ot	TCOME UNCERTAINTY	29	OTHER DECISION PROCESSES
16 16 (16 16	2 determinability 3 high/low uncertainty 4 subsequent acts		291 negotiation

19 OTHER DECISION SUBSTANCE

Table 2-1: SITUATION TAXONOMY — SUMMARY

	USE	R'S OF	PTIONS	3	INP	UT ST	RUCTURE		45	ELIC	CITATION TECHNIQUE
	11	USE I	DECISION ANALYSIS		31	UNCE	RTAINTY				for discrete probabilities
		AT AI	LL?							452	for continuous
							explicit modeling			452	variables for values
			intuitive				time horizon				use group
			decision analysis				subsequent acts			434	elicitation?
		119	other			314	event sequency				elicitation:
							ordering				
							detail level	5	OUT	PUT	
	12	DOLLA	AR AMOUNT OF ANALYSIS			316	degree of grouping		00.		
		121			22	VALU	T. C.		51	SPEC	CIFICATION
			medium high		32	VALU	E				
		123	nign			321	comprehensive?			511	preferred decision
							decomposed?				single value for
	13	POTE	OF DECISION ANALYSIS				partial list?				each option
	13	ROLL	OF DECISION ANALISTS				single index?			513	value distributions
		131	private/public aid				function?			519	other
		132	prescribed/optional			323	runction				
		133	contingent/current								
		433	analysis		33	SPEC	TAL FORMS		52	DISE	LAY FORMAT
		134	optimization/display				Alla Tolaio				
			communication			331	Markov				graphic
							Pareto				computer
							linear programming			529	other
	14	ORGAN	NIZATION								
			analysis source	4	INP	UT SP	PECIFICATION		53	ANAI	LYTIC DEVICES
			input source								
		143	"vest-pocket." relation								use simulation other analytic
			to decision maker?		41	DECI	SION OPTIONS			239	devices
											devices
							specificity of definition				
	15	RESOL	JRCES			419	other	6	MOL	EL MA	ANAGEMENT
									1102		
		151	use a decision		42	EVEN	ITS				
			analysis expert?		42	LVEN			61	MODE	EL DYNAMICS
		152	use a computer?			421	scenarios				
						422	specific			611	combining
							other			612	pooling
2	MOD	EL API	PROACH OPTIONS			,	o cine.				sequential modeling
										614	decision option
					43	VALU	E CRITERIA				scanning
	21	COMP	LEXITY OF ANALYSIS							615	input iteration
			simple			431	units				
			moderately complex			432	base		62		TINGENT ANALYSIS INPUT
		213	very complex			433	evaluation date(s)			SEC	QUENCE
							for time stream				
	22	COMP	REHENSIVE/PARTIAL ANALYSIS								values
	22	COMP	CENENSIVE/PARTIAL ANALISTS								priors likelihood
					44	INDI	RECT ASSESSMENTS				data
		221	comprehensive							024	data
			partial								
						441	conditioned assess-				
						142	ment model? Bayesian updating?				
	23	APPRO	XIMATE ANALYSIS?				decomposed assess-				
						443	ment model?				
		230				444	regression?				
		231	yes				other				

Table 2-2: ANALYTIC TAXONOMY - SUMMARY

above is unnecessary. However, because it is typically very difficult to determine this rating, performance measures must serve an intermediate purpose in matching analytic options to situations.

*

Table 2-3 contains a summary taxonomy of performance measures, and Appendix C presents a detailed description of each measure. In future references, the performance measure numbers will be preceded by the letter "P." For instance, "conceptual completeness" will be referenced as Plll. By contrast to the situation and analysis taxonomies, specific values on the dimensions have not been coded as the final digit.

1	Quality of Decision	3	Other Considerations
11	LOGIC OF CHOICE	31	ACTIVITY PRECEDING CHOICE PROCESSES
	111 conceptual completeness 112 effective disaggregation 113 sound predictions 114 good overall logic 115 scope 119 other		311 good environment monitoring 312 good decision identification 313 good option generation 314 good pre-analysis of anticipated decision
12	QUALITY OF INPUT		319 other
	121 good data gathering 122 good management of staff/ expertise	32	ACTIVITY FOLLOWING CHOICE PROCESSES
	123 posing meaningful questions 124 good overall input quality 129 other		321 good decision communi- cation 322 good hindsight eval-
2	Time and Cost		uation 323 effective implementa- tion
21	ELAPSED TIME		329 other
	211 short elapsed modeling 212 fast input assignment 213 fast calculation 214 fast interpretation 215 short overall net elapsed	33	ORGANIZATIONAL AND OTHER NON-"CHOICE SPECIFIC" IMPACTS 331 improved information
	time 219 other		332 improved command, control, and communication
22	COSTS		333 improved body of applied precepts 339 other
	inexpensive analysis inexpensive input assignment inexpensive calculation inexpensive overall other		

Table 2-3: PERFORMANCE MEASURES - SUMMARY

3.0 MATCHING HIGHLIGHTS

In Section 3.1, we shall generate matching hypotheses about the circumstances that justify the use of any decision analysis based on our applied experience. In doing so, we shall uncover and identify the relevant situation taxonomy that these generalizations imply. Throughout this effort, the emphasis is on ensuring that the generalizations address the proper situation taxonomic considerations rather than on ensuring that the specific generalizations are correct. It is certainly not argued that these generalizations are ready for adoption as they stand as tenets of recommended decision-analytic practice.

There is, in principle, a different situation taxonomy for every level of decision-analytic alternatives (despite a substantial amount of overlap). For example, to decide whether to use decision analysis at all requires characterizing a decision situation according to stakes, reaction time, and the like, whereas deciding whether to use the Delphi method for group elicitation requires such information as whether the informant group has heterogeneous status and homogeneous information.

It is not feasible to itemize exhaustively the union of all such special-purpose sets of characteristics, and it is even less feasible to generate a complete set of matching generalizations for all decision-analytic alternatives and their associated situation taxonomies. To attempt any kind of completeness is to attempt to codify the whole state-of-the-art of decision analysis.

Accordingly, since our aim is more modest, we shall explore only a limited set of matching generalizations that appear either important in their own right or are in some sense representative of a complete taxonomy. The primary motivation is to generate a useful, somewhat general-purpose situation taxonomy that suggests interesting, even provocative, selective generalizations about the state-of-the-art. This preliminary and partial treatment may provide a starting point for developing definitive situation taxonomies and matching generalizations. In Section 3.2, we extend the matching concept to the determination of the form of decision analysis to use.

3.1 Amount of Decision Analysis

In this section, we restrict attention to the most common and perhaps most critical analytic choice: how much

decision analysis, if any, to perform. This choice will be analyzed in two steps. First, decision analysis will be characterized in terms of the performance measures introduced in Section 2.3. Second, the analytic choice of how much decision analysis to perform will be related through performance measures to the situation taxonomy of Section 2.1.

Throughout this section, references to the situation taxonomy will be preceded by the letter "S" and references to the performance measure taxonomy will be preceded by the letter "P." These items are explained in detail in Appendices A and C respectively.

3.1.1 Performance characteristics of decision analysis - The validation of any of the matching generalizations in Sections 3.1.2, 3.2.1, 3.2.2, and 3.2.3 will be through the explicit or implicit mediating role of performance measures summarized in Section 2.3 above.

This section explains how, in general, decision analysis performs along the different performance measures. In all cases, the comparisons are between decision analysis and conventional decision-making practice as exemplified by intuitive or informal decision-making procedures. Decision analysis is not being compared with other quantitative approaches such as operations research or management science techniques although we recognize that these techniques are often appropriate. In all cases, the statements presented are broad generalizations and there are exceptions. The generalizations represent highly personal judgments of the authors, but they are grounded in substantial experience.

The following performance measure generalizations are presented in the order in which the measures appear in Table 2-3, "Performance Measure Summary," rather than in order of importance.

Decision analysis generally performs well in the first performance area, quality of decision (P1). That is, to the extent that this performance measure is important, a large amount of decision analysis is indicated. Under the logic of choice category (P11), decision analysis performs well on most sub-categories. In particular, use of more decision analysis generally leads to: more effective disaggregation of the problem into manageable sub-problems (P112), sounder predictions based upon the available data

Parenthetical references are to the performance measure taxonomy detailed in Appendix C and summarized in Table 3-2.

(P113), and better overall logic (P114). However, it is not clear whether more decision analysis leads to a more conceptually complete consideration (P111). On one hand, using decision analysis tends to increase understanding, but, on the other hand, using decision analysis can lead to serious omissions. Thus, on balance, the use of more decision analysis leads to better overall logic (P114).

In the quality of input category (Pl2), decision analysis performs well on all sub-categories. In particular, the use of more decision analysis will, frequently, promote:

- o Better data gathering (P121);
- o Better management of staff and expertise (P122); and
- o More meaningful questions (P123);

all of which result in better overall input quality (P124).

While the quality of decision (P1) usually increases with an increased use of decision analysis, the time and costs involved (P2) also increase.

First, using decision analysis will generally increase elapsed time (P21) on a first pass, which includes slower first-pass modeling time (P2111), slower first-pass input assignment (P2121), and slower first-pass calculation (P2131). It is not clear whether or not the first-pass interpretation speed (P2141) will be affected by the amount of decision analysis. Thus, overall first-pass elapsed time (P2151) will be longer.

Additional passes of the analysis will, however, generally be faster when decision analysis is used. While it is not clear that this is the case with the elapsed modeling time (P2112), it is the case with input assignment (P2122), calculation (P2132), and interpretation (P2142). Thus, overall time for additional passes (P2152) is reduced. The net for the total overall elapsed time (P2153), including both first and additional passes, is generally greater when decision analysis is used.

In the cost category (P22), more decision analysis always raises the cost of the decision analysis. Greater analytic cost occurs in all of the sub-categories: analysis cost (P221), input assignment cost (P222), calculation cost (P223), and overall cost (224). It occurs for both the first pass and additional passes. It should be emphasized that cost as used in this context refers to the cost of the decision analytic effort, not to gains or costs that might accrue to decision

outcomes from having applied more or less decision analysis to the problem. The trade-off of analytic costs (how much analysis) with the possible impact of that analysis is one of the uses of the proposed taxonomy.

In the third area, other considerations (P3), the effects of decision analysis cannot be generalized for most sub-categories. Using more decision analysis does not have a clearly generalizable effect on environment monitoring (P311), decision identification (P312), option generation (P313), pre-analysis of anticipated decisions (P314), implementation (P323), information processing (P331), or command, control, and communication (P332). However, use of decision analysis will generally improve decision communication (P321), hindsight evaluation (P322), and the body of applied precepts (P333).

The amount of decision analysis indicated in a situation can be determined by a direct assessment of the relative importance of the performance measure. That is, to the extent that the performance measures favoring decision analysis are more important than those not favoring decision analysis, more decision analysis is indicated. However, since it is typically difficult to determine the relative importance of the performance measures directly, it is usually more convenient to use the following situation taxonomy to determine the proper amount of decision analysis.

3.1.2 Situations favoring the use of decision analysis—
The above section explained that a situation characterization in terms of performance measures is sufficient to determine the appropriate amount of decision analysis. However, because this characterization is often difficult to obtain, the situation taxonomy presented in Section 2.1 was developed. This taxonomy is a more natural way to classify situations, but it is not directly relatable to the appropriate amount of decision analysis. 2

This section summarizes the most important aspects of the means by which situations can be related to the appropriate amount of decision analysis through the performance measures. Appendix D contains a more detailed matching of how much decision analysis to use in a situation which addresses all situation dimensions shown in Table 2-1.3 Unless otherwise

²The matching task is conceptually analogous to the formulation of a regression problem. This analogy is presented in Appendix F.

³Situation dimensions are detailed in Appendix A.

noted, situations that support the use of decision analysis also support the use of more decision analysis. (Whenever a statement is made that a certain situation characteristic supports the use of decision analysis, it implies that the more that an actual situation exhibits the situational characteristic, the greater the quantity of decision analysis that is supported.) All statements are guidelines and need to be viewed as a whole rather than in isolation. Combined effects of several dimensions may cause decision analysis to be favored where no single categorization is sufficient to make that indication.

As shown in Table 3-1, a large amount of decision analysis is most strongly recommended in a situation with:

- o Clear options (S1221)4
- o A difficult choice (S1312)
- o A choice as the key consideration (S1334)
- o A maximum option impact greater than \$10 million (S1444)
- o Several substantive and preference sources (S2215)
- o Required pre- and post-decision justifications (S2343)
- o A decision maker who is familiar with decision analysis (S24213).

In addition, a large amount of decision analysis is supported, but less strongly, by a situation with:

- o A current choice (S111)
- o A recurring choice (S1124)
- o A difficult valuation (S1512)
- o Indeterminable uncertainties (S1621)
- o Months of reaction time (S214)
- o Organizational processes resembling a rationalactor model (S2372)
- o Availability of many computational facilities (S2512)
- o A strong staff (S2522)
- o A high degree of decision maker availability (S2542)
- o A high degree of dollar availability (S2563).

An example illustrating the use of these rules is presented in Section 4.1.1 below.

⁴Situation dimensions are detailed in Appendix A.

		SITUA	Amount of Solitsian Analysis	None Lo -<\$15K Med - \$15K-\$50K H1 -> \$50K	
	Basic	S1111 S1124	Current Choice More than two expected # of Occurrences	××	Indicates the key situation characteristics
	Options	S1221	Clear Options	×	e key situat
Decision Substance	Decid- ability	S1312 S1334	Difficult Choice Key Consideration Choice	XX	ion characte
ubstance	Stakes		Maximum Option Impact < \$100K Maximum Option Impact \$100K-\$5M Maximum Option Impact \$5M-\$10M Maximum Option Impact > \$10M	× × ×	ristics
	Outcome Valuation	S1512	Very Difficult Net Valuation	×	
	Outcome Uncertainty	S1621	Indeterminable Uncertainties	×	
	Reaction	S214	Months of Reaction Time	×	
Deci	Analytic Process	\$2215	Several Input Sources for Both Preferences and Substantive Inputs	×	
Decision Processes	Organiza- · ·tional Frocesses	2343 2372	Both Pre and Post Justification Rational Actor Model is Good Approx.	×	
ses	Decision Maker Character- istics	S24213	D. M. Has Much Familiarity with Decision Analysis	×	
	Resources	\$2512 \$2522 \$2542 \$2563	Many Computational Facilities Available Strong Staff Available Much Decision Maker Availability Greater than \$50,000 Available	× × ×	

Table 3-1: MATCHING HIGHLIGHTS - AMOUNT OF DECISION ANALYSIS

The following paragraphs summarize the reasoning behind the primary situation characterizations that support the use of decision analysis.

Decision analysis can improve conceptual completeness (Pll., most when the choice options are clearly specified (S1221).

Decision analysis can provide the greatest improvement in the decision quality (P1) for difficult choices (S1312). When choice is difficult, informal decision processes leave much room for improvement, which can be filled by using decision analysis.

Decision analysis can generally improve choice logic (Pl1). Thus, decision analysis is indicated in situations in which choice is the key consideration (S1334).

The stakes involved in the decision (S14) are the single most important situation classification for determining the appropriate amount of decision analysis. In general, the cost of the analysis (P22) becomes less important as the stakes increase. Thus, decision analysis is indicated in high stakes situations. As a rough rule of thumb, we suggest that the maximum option impact, the dollar equivalent difference in outcome attributable to the choice of the best decision over the worst reasonable decision, should be on the order of \$100,000 (S1442) to justify any decision analysis. Of course, decision analysis is more strongly indicated if the maximum option impact is greater than this threshold.

Conventional, intuitive decision practice has difficulty organizing input from a variety of sources. Decision analysis, however, provides a method for organizing a diverse set of inputs. In particular, decision analysis can improve the quality of both input (Pl2) and choice logic (Pl1) when multiple input sources (S2215) exist. Input quality is especially improved through the effective management of staff and expertise (Pl22), and choice logic is especially improved through effective disaggregation of the decision problem (Pl12).

Decision analysis can provide a vehicle for good decision communication (P321), which is required in situations in which decision justification is needed (S2343).

In a situation in which the decision maker is familiar with decision analysis (S24213), the analysis can provide its greatest benefits, especially the effective implementation of a decision (P323), the effective management of staff expertise (P122), and good overall decision logic (P114).

Secondary considerations that support the use of decision analysis are as follows:

Current choice (Sllll) favors the use of decision analysis because it is certain that the analysis could be used. Since decision analysis is a tool to aid in decision making, its value as an aid is reduced in contingent situations (Slll2) in which it is uncertain that the decision will ever occur.

A decision that is expected to recur (S1124) supports the use of decision analysis because the cost (P22) per analysis is reduced. The amount of advantage gained by recurrence, of course, depends upon both the expected frequency of recurrence and the similarity of the recurring situations because the similarity determines the amount of modification needed before the analysis can be used again. The expected number of occurrences is, in general, closely related to the current/contingent question (S111), explained above, and the two dimensions should be considered simultaneously. For instance, a current decision that will occur only once should be considered the same as a contingent decision that has a 50% chance of occurring twice. Both of these situations have an expected occurrence rate of one.

Decision analysis can improve choice logic (Pl1) in situations in which valuation is difficult (S1512).

Decision analysis can improve choice logic (P11) in situations in which uncertainties are indeterminable (S1621), in the sense that it is difficult to determine what the appropriate probability distribution should be (regardless of whether it has a large or small dispersion).

More time is required to perform a decision analysis than to use informal, intuitive decision-making techniques. Thus, situations with long reaction time (S214) support the use of decision analysis.

Since decision analysis assumes a rational-actor model, a situation that is a good approximation of the model (S2372) supports the use of decision analysis.

Good computational facilities tend, in general, to reduce the cost of analysis (P221). Thus, decision analysis is supported where good computation facilities exist (S2512).

A strong staff (S2522) is most likely to perform a decision analysis correctly and thus improve the overall logic of the choice (Pl14).

If a decision maker is unavailable during the analysis time, the improvement in choice logic (P11) and input quality (P12) provided by decision analysis is reduced. Thus, decision analysis is supported when the decision maker is available (S2542).

When a large resource is available for an analysis (S2563), the cost of the analysis becomes unimportant. Thus, the use of decision analysis is supported by a large available resource.

This section has highlighted the key situation dimensions that support the use of decision analysis. A more thorough examination of the relationship between the situation taxonomy and the amount of indicated decision analysis is presented in Appendix D.

3.2 Examples of Type of Analysis

In this section, the matching concept of Section 3.1 is extended to the determination of the form of decision analysis to use in a given situation. This matching is summarized in Table 3-2. This section will highlight the summary by presenting three examples:

- Situations favoring the use of a contingent choice analysis,
- o Situations favoring the use of a computer to perform the analysis, and
- o Situations favoring the use of the Delphi assessment technique.

Appendix E contains a detailed description of all of the matching generalizations shown in Table 3-2. A case illustration of this part of the taxonomy matching in use appears in Section 4.1.2.

3.2.1 Contingent analysis (Al331)⁵ (analysis in advance of a decision that is expected to arise in the future) - Performance characteristics: A contingent choice analysis requires a large amount of time initially (P2151).

⁵Analytic options are described in Appendix B.

⁶Performance measures are described in Appendix C.

DECISION ANALYSIS CONTINENT CHOICE OPTIMENT CHOICE INT A COMMITTEE	× × × × ×	13 (14) (14) (14) (14) (14) (14) (14) (14)	20123485 000 4577 [EETS \	10 10 10 10 10 10 10 10 10 10 10 10 10 1	NOIN NOIN SOURCE SHOW	10 D D D	10		WINDSHOOM ESTS	148 148 188 188 X		July Nosas Inda	Station Con the name of the State of the Sta	-0,,	1000 1 100 100 100 100 100 100 100 100	OMERCIAL STREET	1987/1845 1845	378V7. 240815 ×	An Ora	COMP 15 18 18 18 18 18 18 18 18 18 18 18 18 18	COMPLETE STREET (623), (124)
VERY COMPLEX ANALYSIS PARTIAL ANALYSIS-IMERENCE ONLY	3	×	×		× ×	× ×		×	×		× × ×	×		×	×	a ×	< ⊠	×	×	X NO JUSTIFICAT	HIGH INFUT LIEBATION (MG1555) NO JUSTIFICATION NEEDED (S2340)
APPROXIMATE ANALYSIS SHORT TIME HORIZON SUBSEQUENT ACTS AS EVENTS			×		×	×	8	×	×								×		×	KANUAL COMPUTATION LOW COMPLEXITY (A23) LOW COMPLEXITY (A23) LOW COMPLEXITY (A23) LOW COMPLEXITY (A23)	7710N (A221) (A221) Subsequent Analysis
DECOMPOSE VALUES ADJUSTED VALUE INDEX					×	×					×	× .		××		×	×		×	NATURE OS VALUE OPTIMIZING MODE	t (A[54])
LINEAR VALUE FUNCTION HARROY MODEL PARETO MODEL			×			×		×				×				×			×	SPECIAL STRUCTURE NEGOTIATION (\$291)	291)
LINEAR PROGRAMING MODEL FLOATING VALUE BASE		8			×	××										i.	××	×		LANGE OFTION VECTOR (S125) CERTAIN (S1650), LINEAN, ST SURTLE TO COMMUNICATE (S254)	ARGE OFFLIN VECTOR (S12%) CERTAIN (S1630), LINEAN, STRUCTURED SUBTLE TO COMMUNICATE (S2540)
MANY CONDITIONING TERS BAYESIAN PROBABILITY MODEL REFERENCE GAMBLE ELICITATION			××		×	×		8			×	×	++-			×	× × ⊠		×	EACH TIER ACCOUNTS FOR OF TOTAL VARIANCE SPECIAL PERCEPTUAL STRU	NACE NOW TOTAL STRUCTURE
AISH DEUM ASSESSMENT ASSISS SIMULATION ASSISS TRULATION ASSISS TEP-THROUGH SIMULATION						×		×			8				+++-		8		×	STATUS HETEROC COMPLEX STRUCT CONDITIONED A	STATUS HETEROGENEOUS, DATA HOMOGENEOUS COMPLEX STRUCTURE (AC13), CONDITIONED ASSESSMENT (AU413)

Table 3.2: MATCHING SUMMARY — TYPE OF ANALYSIS

However, it provides a fast response on subsequent uses (P2152). In addition, in a situation characterized by a short reaction time (S211)⁷, performing a contingency analysis in advance can improve choice logic (P11), input quality (P12), and control (P332).

Situations favoring contingent analysis - The most important situation considerations for determining whether to perform a contingent analysis include the number of occurrences (S112), the stakes (S144), and the reaction time (S21).

In general, contingent choice analysis is indicated in situations in which its cost (P22) is justified by either a large number of occurrences (S1124) or by large stakes (S1444). As a rule of thumb, a contingent analysis should not be performed unless the expected number of occurrences is at least one (P1122) (for example, occurring once with certainty or occurring twice with a 50% probability). It is also necessary to consider the similarity as well as the number of the occurrences. The more similar the situations are, the more the contingent analysis is justified. (As a rule of thumb, three similar recurrences are roughly equivalent to a single identical recurrence.) For a good pre-analysis (P314), it is critically important that the actual decision is predictable, that is, involves the same considerations that are modelled.

As a guideline, the expected option impact over a number of occurrences should exceed \$10 million (S1444) to justify the cost of a contingent analysis. Brown et al. (1975) present an illustration of this guideline in a Navy task force commander's decision situation.

In a situation with a short reaction time (S211), a pre-analysis of contingencies can improve decision quality (P1) by providing a logical framework that considers all available data.

In addition, a contingent analysis is supported, but less strongly, by a situation with clear options (S1221) and several value sources (S2213). Clear options (S1221) allow a contingent analysis to provide a good pre-analysis of the anticipated decision (P314). Clear options also enhance predictability (see above). Contingent choice analyses can improve command, control, and communication (P332) in situations that require several value sources to be considered (S2213).

 $^{^{7}{}m Situation}$ dimensions are described in Appendix A.

3.2.2 Use of computer (Al521) - Performance characteristics: A computerized analysis is generally time-consuming and costly to set up (P2131 and P2231) but is fast and inexpensive to run (P2132 and P2232). However, when an analysis is complex, a computer may not increase the initial time or cost.

Situations favoring the use of a computer include a recurring situation (S1124) or a situation with large stakes (S1444). Since a computer offers faster and less expensive performance on subsequent passes, its use is supported by a large expected number of occurrences (S1124). Large stakes argue for the use of a computer for two reasons. First, large stakes justify the large costs needed for a computerized analysis. Second, a situation with large stakes generally involves a complex analysis. In this case, a computerized analysis may not be more expensive than a manual analysis.

The following situations support the use of a computer, but less strongly than those mentioned above.

A computer analysis requires at least several days of anticipatory reaction time (S213). However, since a computerized analysis runs quickly, only minutes of execution reaction time (S211) are needed.

Some staff support (S2521) is needed to program a computerized analysis.

The costs (P22) and time (P21) of a computerized analysis are not significantly worse than the alternative of a non-computerized analysis if the situation has clear, complex options (S1221 and S1236); many determinable uncertainties (S1613 and S1623); and many measurable values (S1523 and S1532).

A computer is also recommended in conjunction with a complex analysis (A213) and with simulation (A5311).

3.2.3 The Delphi Technique (A45411) (a group elicitation technique for probabilities) - Performance characteristics: The Delphi technique involves pooling opinions of several probability assessors while allowing only limited interaction among them.

Situations favoring the use of the Delphi Technique include only those with several uncertainty sources (S2214).

The Delphi technique limits interaction among assessors and provides the assessors with anonymity. Thus, the Delphi technique removes inhibition and allows better use of staff expertise (Pl22) in cases in which the status of the assessors is heterogeneous. However, if data is heterogeneous, then the Delphi technique reduces input quality (Pl2) by inhibiting useful interactions. (See Fischer [1975] for a further evaluation of the Delphi technique.)

4.0 MODE OF USE

The primary objective of the taxonomy of decision situations presented above is to provide the decision analysis user with a convenient device for recognizing promising situations in which to use decision analysis, and to guide him in making the kinds of analytic choices illustrated in Sections 3.1 and 3.2. At the current state of development, the situation taxonomy, though incomplete, is too complex to be applied routinely in unfolding decision situations.

What is ultimately needed is a procedure for progressively applying the taxonomy so that it will operate like a succession of progressively finer screens. The answers to a few routinely applied questions, perhaps related to the size of the stakes involved, the time available to make a decision, the clarity of the choices, and the complexity of the outcomes, will signal which situations should be seriously considered for decision analysis. For those, a second level of more sharply focused questions will be used to make a definitive decision on whether to use decision analysis and, if so, what role the analysis will serve. After resolving these issues, even more sharply focused questions will be used to determine the detailed planning of the analysis. The screening would continue until the exact analytic options are identified, or until informal technical judgment is sufficient to designate the type of analysis.

4.1 Application of Universal Taxonomy

As explained in Section 1.2, this paper has developed a universal situation taxonomy to aid in matching decision—analytic options to decision situations. Section 4.1.1 below illustrates how the universal taxonomy should be applied to a decision situation in order to determine the appropriate amount of decision analysis, and Section 4.1.2 illustrates the use of the universal taxonomy in choosing the type of analysis for the same situation.

4.1.1 Amount of decision analysis to use on export control decision - In the summer of 1973, a senior staff member of the President's Council of International Economic Policy (CIEP) had to make a recommendation to the President on the level of embargo for computers sold to the Soviet Bloc. If approved by the President, in consultation with the Departments of State and Defense, this recommendation would provide the basis for the U.S. position on COCOM embargo policy. COCOM, an international coordinating committee representing the major western powers and Japan, is charged

with controlling the export of computers and other strategic items to communist countries. The essential decision by COCOM was to specify where, in terms of computing power, to set the "easy access line" for computers.

Six weeks before his recommendation was due, the CIEP staffer was faced with a very large array of reports from experts on the complex issues involved in the decision. These issues included the impact of the decision on military threat, U.S. computer sales, the attitude of COCOM allies, and other economic and political considerations. Although the staffer had some sense of where the "easy access line" should be set, he felt uneasy about his ability to properly digest and interpret the complex mass of data available to him within the short time available.

The first question, then, was how much, if any, decision analysis the staffer should have done, and, secondly, what specific form the decision analysis should have taken.

Table 4-1 indicates how the matching highlights shown on Table 3-1 might be used, informally, to conclude that a substantial amount of decision analysis should be used. It can be seen that all but three of the seventeen situation characteristics favoring a large amount of decision analysis are present in this case, including all but one of seven key characteristics.

Specifically, proceeding from left to right, we see that the choice was a current one, and that the decision situation is expected to recur (the issue is expected to come up for re-examination every two years or so). options are somewhat clear, in the sense that the theoretically rich and complex option space could be reduced to six specific alternatives without gross over-simplification. The choice was, however, not particularly difficult in that the decision maker was rather sure that the right choice lay between three of the six options. The decision maker saw his key problem as choosing from among perceived options (rather than making predictions or determining what the options are, for instance). All measures of stakes were very high, the maximum option impact was on the order of hundreds of millions of Federal budget dollars. The net valuation of possible outcomes was very difficult (trade-offs of economic, political and military considerations were required), but there was no unusually great indeterminability of uncertainties. The reaction time was on the order of a couple of months. There

		SITUA'	Amount of Amount of Amount of Decision Analysis	None Lo- \$15k Med-\$15k-\$50 H1- \$50k	
	Basic	S111! S1124	Current Choice More than 2 expected # of Occurrences	× ×	
	Options	S1221	Clear Options	· ×	
Decision Substance	Decide- ability	S1312 S1334	Difficult Choice Key Consideration Choice	× ×	
Substance	Stakes	S1441 S1442 S1443 S1444		× × ×	
	Outcome Valuation	\$1512	Very Difficult Net Valuation	×	
	Outcome Uncertainty	S1621	Indeterminable Uncertainties	×	
	Reaction	S214	Month of Reaction Time	×	
Dec	Analytic Processes	S2215	Several Input Sources for Both Preferences and Substantive Inputs	×	
Decision Process	Organiza- tional Processes	2343 2372	Both Pre and Post Justification Rational Actor Model is Good Approx.	×	
s	Decision Maker Character- Istics	S24213	D. M. Has Much Familiarity with Decision Analysis	×	
	Resources	\$2512 \$2522 \$2542 \$2563	Many Computational Facilities Available Strong Staff Available Much Decision Maker Availability Greater than \$50,000 Available	× × × ×	

+ Indicates characteristic is present - Indicates characteristic is not present

Export Control Characterization

‡

Table 4-1: EXPORT CONTROL EXAMPLE - AMOUNT OF DECISION ANALYSIS

were several different sources of input for the various types of uncertainty and prediction, all of which were different from the sources of value judgment. The decision required justification in a formal report to the President and in verbal presentations to representatives of industry and government agencies both before and after the decision was to be made. The decision processes of government for this decision are reasonably "rational" (in that reasoned argument is a large part of the persuasion process). decision maker was highly familiar with decision analysis (having taught the subject at a major univeristy). Computational facilities and staff skills were more than adequate. The decision maker was able to spend large amounts of time participating in the analysis on a regular basis. However, the budget available was quite modest, on the order of \$30,000.

At this stage of taxonomy development, no formal decision rule is proposed for determining routinely what level of decision analysis should be used. In this case, however, it is clear that any plausible decision rule would indicate a large amount of decision analysis, since only one of the seven key situation characteristics is absent (difficult choice), and only two of the other ten favorable situation characteristics are absent (indeterminable uncertainty and large budget). However, since only a small budget (\$30,000) for analysis was available, its size alone easily determined the dimensions of the project without any need for a formal decision rule.

4.1.2 Type of decision analysis to use on export control decision - The following discussion indicates the appropriate type of analysis for the export control problem. This discussion, which treats each analytic option, should be considered in conjunction with Table 3-2 above.

The analysis was used in a display rather than in an optimization mode (A1342) because the scope of the analysis, at the client's request, was not comprehensive (A222); certain political dimensions of value such as impact on detente were deliberately excluded from the analysis. The display of the decision's impact on those dimensions of value that were explicitly modeled (economic, military, and so forth), however, permitted an evaluation of how large the excluded political considerations would need to be in order to change the preferred option.

Watson and Brown (1975/1) presents an attempt to put a value on this particular analysis, after-the-fact.

A computer was used (Al521) because the decision was expected to recur about every two years (Sl124), the stakes were high (Sl444), and many input iterations (A61533) were expected. The large number of input iterations was indicated because this was a very controversial decision (S2352).

The complexity of the analysis was low (A211) because of a need to communicate (P321) the arguments in order to justify the decision (S2343) and because few of the conditions indicating complexity were met (see Table 3-2). In particular, only one probability distribution was modeled (on one dimension of value--military threat).

Subsequent acts were not explicitly modeled (A3130) because they did not have enough impact (S1641) to justify any explicit modeling.

On the other hand, value was decomposed extensively (A3222) into its four major dimensions (U.S. computer sales, allied goodwill, military threat, and other) and one dimension, military threat, was decomposed into sixteen specific variants. One reason was that the sources of preference judgment were different (S2213) and partly because the decision was controversial (S2352), especially regarding trade-offs among value dimensions (for example, how much improvement in computer sales will compensate for a deteriorating military threat.)

The decompositions were all linear (A32511) because of a need for both pre- and post-justifications (S2343).

The value scales used a "floating base" (A4322) because the technical problems of valuation of each dimension were not very difficult (S1510) and because experienced staff (S2522) were available to handle the greater technical subtlety involved.

In performing the calculations, simulation was not needed (A5310) because the model structure was not complex (A211).

4.1.3 Sequential taxonomy matching - Direct use of the universal taxonomy, as illustrated above, may present substantial problems of elicitation if all potentially relevant characteristics are involved. Since there are hundreds of situation characteristics to measure, it may be difficult to measure any one of them, let alone all of them. That is, it may be very difficult to locate precisely any given situation in situation space.

However, for any particular choice (for example, whether to use decision analysis at all), only some of the situation dimensions are relevant and even all of those may not need to be evaluated to make a given analytic choice. the choice of whether to use decision analysis or not may be resolved on the grounds of stakes alone if they are small enough. For example, the small size of the budget (S2562) and the large size of the stakes (S1444) in the COCOM Case (see Section 4.1.1) immediately determined the amount of analysis.

Thus, a sequential elicitation of situation characteristics is often indicated. In principle, a flow chart could be constructed to control the sequence of situation characteristic elicitation. The answers to each situation question either indicate an analytic choice or call up a new question. Figure 4-1 shows notionally the form such a flow chart might take, but it does not represent a considered proposal. A complete flow chart of this form would represent, in effect, a manual of applied decision analysis.

4.2 Indirect Taxonomy Matching Through a Specific Taxonomy

Since relevant situational characterizations tend to be difficult to accomplish, it is desirable to seek mediating characterizations that enable an easy characterization to serve as a surrogate for a relevant one. For this reason, specific taxonomies, which are specific to a particular field of decision-making and which are based upon concepts that are familiar in the field, need to be developed in addition to the universal taxonomy presented above, which is not specific to a particular area of decision making.

Figure 4-2 illustrates the manner in which a situation can be characterized along both specific and universal taxonomies in order to determine the appropriate decision—analytic techniques. The decision to neutralize an enemy—held island, which may cause a global war, might be characterized as a "crisis" situation on the specific taxonomy. This characterization is associated with the universal dimensions of "multiple value source," "controversial," and "high stakes." Through matching generalizations, the situation characterization indicates that a substantial scale of decision analysis, requiring explicit disaggregation of value, is indicated. In conjunction with other specific or universal descriptions of the situation, this matching indicates specific analytic choices, step-through simulation model structure, and decomposed valuation.

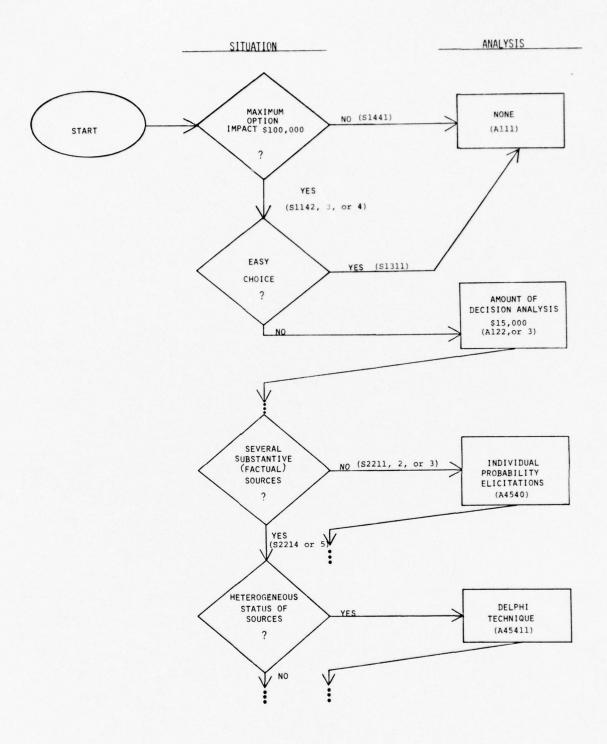


Figure 4-1
ELICITATION SEQUENCE FLOW CHART

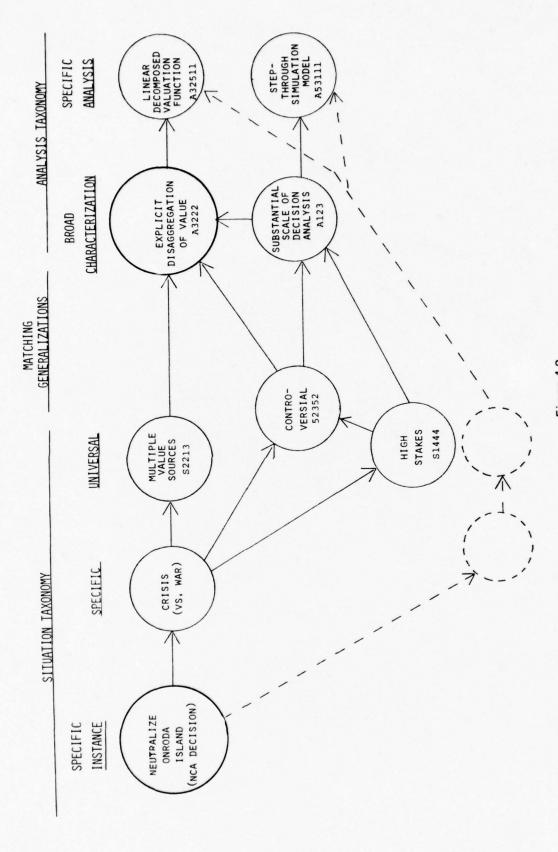


Figure 4-2 ILLUSTRATIVE USE OF TAXONOMY MATCHING

Table 4-2 illustrates another example of how a specific taxonomy matching might be developed. The specific characteristic of business-versus-government decisions is analyzed to determine how much decision analysis to use. Table 4-2 indicates that government decisions are more frequently associated with the universal situation characteristics favoring decision analysis than are business decisions. Table 4-2 suggests, therefore, that, as a general rule (modified of course by particular circumstances, that is, other specific or universal situation characteristics), government decisions favor decision analysis more than business decisions do.

The specific taxonomy examples presented above are intended only to present the idea and point the direction for future research. 2

²In addition, a start has been made in Sections 1, 2, and 6 of Brown et al. (1974) on developing a specific taxonomy for naval tactical situations.

Universal Taxonomy

Characteristics Favoring DA

	Specific Chara	acteristics
	Government	Business
High expected number of occurrences (S1124)		х
Difficult choice (S1312)	х	
High stakes (S14)	х	
Difficult net valuation (S1512)	х	
Several sources for both substantiv and preference inputs (S2215)	e X	
Controversial (S2352)	х	
Need for post-decision justificatio (S2343)	n X	

Specific matching conclusion: government decisions tend to favor the use of decision analysis more than business decisions.

Table 4-2: APPLICATION OF SPECIFIC TAXONOMY

5.0 CONCLUSIONS

5.1 Implications of Work Completed

It is important not to confuse the grand design of the total taxonomy matching task with the modest and tentative beginning which this report represents. What is proposed here is no more than a conceptual framework with some notional substance. Even at its completion, the taxonomy matching will represent essentially a vehicle for communicating the state-of-the-art of applying decision analysis and will make only incidental contributions to the state-of-the-art itself. Possibly the structure developed will encourage methodological inquiry simply by providing a convenient vehicle for dissemination of the results of such inquiry. The main value of the taxonomy, however, must lie in being a prompter, a pointer, and, to some extent, a stage setter for further, more substantive research.

However, this report does carry some of the tentative methodological doctrine that a few specific decision analysts at Decisions and Designs, Incorporated have developed. That doctrine, cast in this format, may stimulate some controversy or methodological inquiry especially when it conflicts with the textbook paradigms of decision analysis (for example, the modeling of subsequent acts as events). However, the proper vehicle for research aimed at doctrinal persuasion, at least in the first instance, is in topic oriented technical reports, not here.

5.2 Future Research

Since this project is only a tentative first step toward the codification of decision analysis, we are well aware that much remains to be done to arrive at something even approximating a definitive framework for ourselves and others in the field to work with. Toward this end, we are soliciting comments and suggestions from colleagues and interested researchers. If responses are sufficiently encouraging, it is planned to modify this material and ready it for general publication.

The following are some proposals for change or development under current consideration:

- Re-order items in the performance measure taxonomy by putting the "overall" categories first;
- Include "Don't Know" and "Not Modeled" categories in the taxonomies;
- Replace the last (response) digit in the taxonomies with a letter. Thus, "binary options" might be referenced as "S123A" instead of "S1231";

- 4. Expand Section A45, elicitation techniques;
- 5. Resolve a possible conflict concerning Pl12 and Pl23: Is the performance measure, "effective disaggregation," meaningful, or is it the same as "posing meaningful questions?";
- 6. Review terminology to be used; for example, should "substantive" or "uncertainty," describe a type of input in S221?; and
- 7. Expand the bibliography.

From this tentative beginning, a number of promising avenues for research appear, including the following:

- Refine and enrich the taxonomic and matching material in this report in the light of more thorough investigation of the same issues, and of a wider range of decision-analytic experience (involving organizations other than DDI);
- 2. Develop specific taxonomies and matching generalizations for particular purposes, for example, a matching taxonomy to select decision aids by task force commanders for tactical naval battle decision situations. This development could be a valuable adjunct to an ongoing program sponsored by the Office of Naval Research on Operational Decision Aids. This particular problem is discussed in Appendix A of Payne et al. (1974) and Sections 1, 2, and 6 of Brown et al. (1974);
- 3. Develop an analogous analysis to that presented here directed to the problem of matching decision situations with information needs (as opposed to analytic approaches). Such an inquiry, or a specific version that is specific to Naval tactics, might be a valuable adjunct to the development of the information component of a tactical flag command center (TFCC) of the type currently under consideration in the U.S. Navy;
- 4. Extend generalizations proposed here to the point of proposing specific decision rules which go beyond the simple listing of favoring characteristics;
- 5. Codify the state-of-the-art decision analysis with respect to specific technical issues, one at a time. This codification would call for a richer and more definitive expansion of sub-sections of Section 3.2, for example, expanding an individual row of Table 3-2;

6. Develop an implementation procedure for the taxonomy matching, for example, by proposing and trying out in practice specific procedures to help a decision maker or a technician evaluate candidate decisions for decision analysis (e.g., by developing a filtering procedure along the lines of Figure 4-1); and,

3

Develop interactive computer graphic programs to streamline the filtering procedure referenced in 6 above.

APPENDIX A

SITUATION TAXONOMY

					1	
	1	Deci	sion Substance	2	Deci	sion Process
	•	2001				
	11	EASI	C SITUATION	21	REAC	TION TIME
		111	current/contingent choice		211	minutes
		112	expected # of occurrences		212	hours
		(113	operating/information act)		213	days
		119	operating/information act) other basic situation		214	months
	12	OPTI	one	22	ANAT	YTIC PROCESSES
	12	OPTI	ONS	22		
			broad/narrow)		221	# of input sources
		122			222	analytic team
		123	complexity of decision options		223	constraints on analytic method
-		124			224	documentation
		125	radical/adaptive static/dynamic		229	other
			other options			Center
				23	ORGA	NIZATIONAL PROCESSES
	13	DECI	DABILITY		(231	initiation)
		121	diffigulty of aboles		232	
			difficulty of choice unfamiliarity		(233	coordination)
			key considerations			justification
			other decidability			controversiality
		134	other decidability		236	
					237	
						risk attitude
	14	STAK	KES		239	other organizational
						processes
		141	resources committed			
			cost swing			
			value swing			ISION-MAKER CHARACTERISTIC
			maximum option impact	24	DEC.	ISION-MAKER CHARACTERISTIC
			expected irrationality cost		241	role in organization
		149	other stakes		242	personal characteristics
						other
	15	OUTC	COME VALUATION			
	13					
		151	difficulty of net valuation # of value dimensions)	25	RES	OURCES AVAILABLE
			measurable value?)		251	computational facilities
		(154			252	staff
		(134	values)		253	decision analysis expert
		(155			254	
		(156				maker
		,130	valuation)		(255	availability of assessors)
		159			256	
					259	other resources
	16	OUTC	OME UNCERTAINTY	29	OTHE	ER DECISION PROCESSES
		161	# of uncertainties		291	negotiation
		162				
			high/low uncertainty			
		164	subsequent acts			
		(165	type of evidence)			
		166	hindsight monitoring			
			other outcome uncertainty			

19 OTHER DECISION SUBSTANCE

*

Table A-1: SITUATION TAXONOMY - SUMMARY

1 Decision Substance

8

The following dimensions describe the substance of a decision as opposed to the setting in which the decision is to be made.

11 Basic Situation

- 111 Current/contingent choice Does the situation demand a choice which will result in a current or contingent action?
 - 1111 Current e.g., mobilize NATO forces
 immediately.
 - 1112 Contingent e.g., mobilize NATO forces if and when the probability of Warsaw Pact forces mobilizing exceeds .55.
- 112 Expected number of occurrences a probability weighted average of the number of times that the decision must be made (including the first time).
 - 1121 Expected number of occurrences < 1
 - 1122 Expected number of occurrences = 1
 (e.g., a current choice or a contingent choice that will occur twice if it occurs at all and has a 50% chance of occurring).
 - 1123 Expected number of occurrences = 2
 - 1124 Expected number of occurrences > 2
- 113 Operating/information act Will the decision result in taking an operating act or an act to seek information?
 - 1131 Operating e.g., shoot at an unidentified plane.
 - 1132 Information e.g., seek information on an unidentified plane's identity.
- 119 Other basic situations.

12 Options

- 121 Broad/Narrow Is a commitment required at a broad or narrow level? (A narrow level is typically a subset of a broad one.)
 - 1211 Broad e.g., go North or go south.
 - 1212 Narrow e.g., go to Fairbanks, Alaska, or go to Miami, Florida.
- 122 Clear/Fuzzy Are the options clearly specified or not?
 - 1221 Clear e.g., selecting one of five bidding contractors.
 - 1222 Fuzzy e.g., selecting criteria to use in selecting a contractor.
- 123 Complexity of decision options How many decision options are under consideration?
 - 1231 Two (a binary choice).
 - 1232 Three to twelve.
 - 1233 More than twelve (discrete).
 - 1234 One scalar e.g., what price to change.
 - 1235 Small vector e.g., what values to assign to three design parameters.
 - 1236 Large vector.
 - 1239 Other
- 124 Radical/adaptive Is the decision within the realm of the decision maker's current practice?
 - 1241 Radical outside of the realm of current practice and the alternatives are very different, (e.g., a decision to drop a product).
 - 1242 Adaptive involving minor changes in current practice and the alternatives are very similar (e.g., a decision to raise the price of a product).

- 125 Static/dynamic Is this decision final or can it be modified over time?
 - 1251 Static The decision is to be made now once and for all (e.g., a declaration of war).
 - 1252 Dynamic The commitment can be modified, possibly continuously (e.g., taking an increasingly bellicose stance toward a potential enemy, maneuvering a ship).
 - 129 Other options

13 Ease of Decision

- 131 Difficulty of choice Is there a lot of room for improvement over informal decision making?
 - 1311 Routine or obvious Informal techniques leave little room for improvement.
 - 1312 Difficult There is much room for improvement over informal decision-making techniques. Decision maker is perplexed about what to do and there is a high "cost of confusion."
- 132 Unfamiliarity How foreign is the choice to the decision maker's experience?
 - 1321 Low familiar to the decision maker's experience.
 - 1322 High foreign to the decision maker's experience.
- 133 Key consideration in this case, what is the key determinant of a good decision?
 - 1331 Option generation generating a set of viable options.
 - 1332 Information gathering information to support a choice.
 - 1333 Inference drawing inferences from available information.
 - 1334 Choice choosing from among a set of options.

1339 Other considerations.

139 Other decidability

14 Stakes

¥

141 Resources committed - What total amount of resources are involved with the decision (e.g., value of facilities engaged)?

1411 Low - less than \$200,000.

1412 Medium - \$200,000 to \$2 million.

1413 High - greater than \$2 million.

142 Cost swing - What is the difference in cost between the least expensive and most expensive option?

1421 Low - less than \$100,000.

1422 Medium - \$100,000 to \$1 million.

1423 High - greater than \$1 million.

143 Value swing - What is the difference in value between the best and worst plausible outcomes (regardless of option)?

1431 - Less than \$1 million.

1432 - \$1 million to \$40 million.

1433 - \$40 million to \$100 million.

1434 - Greater than \$100 million.

144 Maximum option impact - What is the dollar equivalent difference in outcome attributable to the choice of best decision over worst reasonable decision?

1441 - less than \$100,000.

1442 - \$100,000 to \$5 million.

1443 - \$5 million to \$10 million.

1444 - Greater than \$10 million.

145 Expected option impact - What is the expected value of the differences attributable to option choice?

1451 - less than \$50,000.

1452 - \$50,000 to \$2 million.

1453 - \$2 million to \$5 million.

1454 - greater than \$5 million.

146 Expected irrationality cost - What is the room for improvement in the user's decision making process (the expected cost of choosing an option other than that indicated by an "impeccable" decision analysis)?

1461 Low - less than \$150,000.

1462 Medium - \$150,000 to \$500,000.

1463 High - greater than \$500,000.

149 Other stakes

15 Outcome Valuation

151 Difficulty of net valuation - How difficult is it to compare the attractiveness of possible outcomes? (Specific elements of this characterization are covered below in items \$152 - \$155.)

1510 Not very - e.g., for a single criterion with a natural metric.

1511 Moderately.

1512 Very - e.g., for multiple conflicting intangible criteria.

Number of value dimensions - Is a single value criterion (measure of effectiveness) sufficient or are multiple criteria important?

1521 Single

1522 Few

1523 Many

- 153 Measurable value To what extent does a natural metric exist for the value dimensions?
 - 1530 No natural metric e.g., for quality of life.
 - 1531 Approximate metric e.g., GNP as a metric for standard of living.
 - 1532 Exact metric e.g., dollars as a metric of income.
- 154 Natural combinability of value How natural is it to combine the dimensions of value into a single index?
 - 1541 Low e.g., in the case of many conflicting intangible variables.
 - 1542 Moderate.
 - 1543 High e.g., for different additive components of a single variable such as cost.
 - 1549 Other e.g., high for some but not for all.
- 155 Timing early/late Are the value dimensions characterized by early or late evaluation dates?
 - 1551 Early e.g., cost of a contract.
 - 1552 Late e.g., long-run impact on U. S. military standing.
- 156 Difficulty of component valuation How difficult is it to value outcomes on each criterion scale?
 - 1561 Low
 - 1562 Medium
 - 1563 High
- 159 Other outcome valuations.

16 Outcome Uncertainty

161 Number of uncertainties - How many uncertain quantities influence the decision?

1611 One

1612 Few

1613 Many

162 Determinability - How difficult is it to determine the appropriate probability distribution(s) of the uncertain quantities?

1621 Low

1622 Medium

1623 High

163 High/low uncertainty - What is the "amount" of uncertainty in the uncertain quantities (e.g., as represented in the "spread" of the probability distribution)?

1630 None

1631 Low

1632 Medium

1633 High

164 Subsequent acts - To what degree do subsequent acts impact the decision outcome?

1641 Low

1642 Medium

1643 High

165 Type of evidence - What is the nature of evidence available for resolution of uncertainty?

1651 Historical record.

1652 Direct judgment only.

1653 Indirect judgment - e.g., based on conditioning events.

- Hindsight monitoring How accurately and promptly can the outcomes be determined by hindsight?
 - 1661 Low e.g., having to wait years to know whether there were mines at the entrance of the Dardanelles in 1914.
 - 1662 Medium.
 - 1663 High e.g., discovering within hours that a bombing mission was successful.
- 169 Other outcome uncertainties
- 19 Other Decision Substance

2 Decision Process

In terms of the user's analytic options (as opposed to the technician's options), the organizational and personal setting within which a decision is to be made may be of even greater importance than the specific substance of the decision. Thus, we may be interested in the personal characteristics of the decider, the organizational setting within which he operates, or external constraints on the decision process.

- 21 Reaction Time How much reaction time is available between the time that the decision is recognized and the time that a commitment is required?
 - 211 Minutes e.g., in deciding to shoot an approaching unidentified plane.
 - 212 Hours e.g., in the Pueblo incident.
 - 213 Days e.g., in a NATO mobilization decision.
 - 214 Months e.g., in planning a reconnaissance system.

22 Analytic Process

- 221 Number of input sources How many sources will provide input data?
 - 2211 Single for all inputs All inputs will come from a single source.
 - 2212 Single for each input There will be several input sources, but only one source for each input.
 - 2213 Several preference sources There will be several (e.g., preference value) input sources but only a single source for substantive input.
 - 2214 Several substantive sources There will be several sources for substantive (e.g., uncertainty) input but only a single source for value input.
 - 2215 Several for both There will be several substantive and preference sources.

222 Analytic Team

- 2221 Number of team members How many people will actually perform the analysis?
 - 22211 Single e.g., the decision maker himself, a member of his staff, or an outside consultant.
 - 22212 Multiple e.g., a group of staff members or a horizontal task force comprised of several people of equal rank in the organization.
- 2222 Type of analyst Who will actually perform the analysis?
 - 22221 The decision maker himself.
 - 22222 A member of the decision maker's staff.
 - 22223 An outside consultant.
- 2229 Other distinctions.
- 223 Constraints on analytical method What established procedures constrain the analytic method?
 - 2230 None The analysis is unconstrained.
 - 2231 Required The analysis is required to conform to established procedures.
 - 2239 Other
- 224 Documentation What procedures constrain reporting and documentation?
 - 2240 None
 - 2241 Required A certain form of documentation is required.
 - 2249 Other
- Organizational Process These dimensions deal with the way in which a particular decision is articulated within the organization and the way the decision is initiated, used, and controlled.

231 Initiation

- 2311 How is the analysis initiated?
 - 23111 Spontaneously in response to unanticipated developments.
 - 23112 Scheduled ahead of time.
 - 23119 Other
- 2312 Who initiates the decision?
 - 23121 The decision maker
 - 23122 His staff
 - 23123 His superior
 - 23124 A subordinate
 - 23125 A peer
 - 23129 Other
- 232 Responsibility Where, in the organization, is the responsibility for the decision?
 - 2321 Role of decision analysis user What role does the decision analysis user serve in the decision process?
 - 23211 Firm decision He will make the final decision.
 - 23212 Tentative decision He will make a decision that is subject to someone else's revision.
 - 23213 Recommendation He will make a recommendation.
 - 23214 Information He will provide information for someone else's decision.
 - 23219 Other

2322 Dispersion of decision responsibility - How is the responsibility for the decision dispersed?

23220 Not

8

23221 Horizontally

23222 Vertically

23229 Other

233 Coordination - To what degree must the decision coordinate with other higher, lower, or collateral decisions?

2330 Not at all

2331 Moderately

2332 Very much

234 Justification - Is the analysis required as justification before or after a decision is made?

2340 No

2341 Pre only

2342 Post only

2343 Both pre and post

235 Controversial - How controversial is the decision?

2350 Not

2351 Moderately

2352 Very

236 Performance control - what type of accountability, evaluation, or reward system is in operation?

2360 None - The decision maker is his own master.

2361 Loose

- 2362 Tight The decision maker operates under very strict control.
- 2369 Other e.g., the decision maker is controlled by more than one superior (as with the Chief of Naval Operations).
- 237 Rational-actor model How well is the organizational decision process approximated by a rational actor model?
 - 2371 Poor approximation.
 - 2372 Good approximation.
- 238 Risk Attitude Does the process require the consideration of attitude toward risk?
 - 2380 Risk neutral.
 - 2381 Risk averse.
 - 2382 Risk seeking.
 - 2389 Other non-linear utility function
- 239 Other organizational processes
- 24 Decision Maker Characteristics

What are the characteristics of the person making the decision?

- 241 Position in organization
 - 2411 Authority level What authority does the decision maker command?
 - 24111 Low first-line manager (e.g., Lieutenant).
 - 24112 Medium
 - 24113 High top management (e.g., General Officer).
 - 2419 Other e.g., What type of organization does the decision maker belong to?
- 242 Personal characteristics

2421 Familiarity with decision analysis - How familiar is the decision maker with decision-analytic techniques?

24211 None

24212 Little

24213 Much

2429 Other, e.g., management style.

249 Other characteristics

25 Resources Available

Œ.

251 Computational facilities - How much computational facility is available for analysis?

2510 None

2511 Some - small computer.

2512 Much - powerful computer.

252 Staff - What type of staff support is available to the decision maker?

2520 None

2521 Modest - a few or untrained people.

2522 Strong - many competent people.

253 Decision-analytic expert - Does the decision maker have access to decision-analytic expertise either in-house or through consultants?

2530 None

2531 Low

2532 High

254 Decision-maker availability - How available will the decision maker be throughout the course of the analysis?

2540 Not

2541 Little

2542 Much

255 Availability of assessors - How available will the assessors be throughout the course of an analysis?

2550 Not

2551 Low

2552 High

256 Dollars available - How much money is available for the analysis?

2561 Less than \$15,000.

2562 \$15,000 to \$50,000.

2563 Greater than \$50,000.

259 Other resources

29 Other Decision Processes

291 Negotiation - Are there two or more actors with conflicting interests?

2910 No - single actor

2911 Two

2912 More

APPENDIX B

ANALYTIC OPTION TAXONOMY

USER'S	OPTIONS	3	INP	UT ST	RUCTURE		45	ELIC	ITATION TECHNIQUE
								451	for discrete
11 USI	E DECISION ANALYSIS		31	UNCE	RTAINTY			452	probabilities for continuous
AT	ALL?							452	variables
					explicit modeling time horizon			453	for values
	l intuitive				subsequent acts			454	use group
	decision analysis other				event sequency				elicitation?
111	ocher			214	ordering				
				315	detail level				
12 DOI	LLAR AMOUNT OF ANALYSIS			316	degree of grouping	5	OUT	PUT	
	121 1ow		32 VALUE			51		SPEC	IFICATION
	2 medium		32	VALU	Е				
12	3 high			321	comprehensive?			511	preferred decision
					decomposed?			512	single value for
13 PO	LE OF DECISION ANALYSIS				partial list?				each option
13 RO	BE OF DECISION ANALISTS				single index?				value distributions
13	l private/public aid				function?			519	other
	2 prescribed/optional								
	3 contingent/current						5.3	DICE	LAY FORMAT
	analysis		33	SPEC	IAL FORMS		52	DISP	THI LOKUMI
	4 optimization/display			222	Markov			521	graphic
13	5 communication				Pareto			522	computer
					linear programming				other
14 OR	GANIZATION			223	Tricar programming				
14	l analysis sourse		TND	tim Sp	ECIFICATION		53	ANAL	YTIC DEVICES
14	l analysis source 2 input source	4	TIVE	01 31	ECTITION				
14	3 "vest-pocket" relation								use simulation
	to decision-maker?		41	DECI	SION OPTIONS			539	other analytic devices
				411	specificity of definition				
15 RESOL	SOURCES			419	other		MOD	DT MX	NAGEMENT
							MOL	EL PIN	THANKI
15	l use a decision		42	EVEN	ITS				
	analysis expert?								L DYNAMICS
15	2 use a computer?						61	MODE	
					scenarios		61		
				422	specific		61	611	combining
MODEL	APPROACH OPTIONS			422			61	611 612	combining pooling
MODEL	APPROACH OPTIONS			422	specific		61	611 612 613	combining
MODEL	APPROACH OPTIONS		42	422 429	specific other		61	611 612 613	combining pooling sequential modeling
	APPROACH OPTIONS MPLEXITY OF ANALYSIS		43	422 429	specific		61	611 612 613 614	combining pooling sequential modeling decision option
21 CO			43	422 429 VALU	specific other			611 612 613 614 615	combining pooling sequential modeling decision option scanning input iteration
21 CO 21 21 21	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex		43	422 429 VALU	specific other E CRITERIA units			611 612 613 614 615	combining pooling sequential modeling decision option scanning input iteration CINGENT ANALYSIS INPU
21 CO 21 21 21	MPLEXITY OF ANALYSIS 1 simple		43	422 429 VALU 431 432	specific other			611 612 613 614 615	combining pooling sequential modeling decision option scanning input iteration
21 CO 21 21 21	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex		43	422 429 VALU 431 432	specific other E CRITERIA units base			611 612 613 614 615	combining pooling sequential modeling decision option scanning input iteration PINGENT ANALYSIS INPUT DUENCE
21 CO 21 21 21 21	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex		43	422 429 VALU 431 432	specific other E CRITERIA units base evaluation date(s)			611 612 613 614 615 CONT SEQ	combining pooling sequential modeling decision option scanning input iteration PINGENT ANALYSIS INPUT OUENCE values
21 CO 21 21 21 21	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex			422 429 VALU 431 432 433	specific other E CRITERIA units base evaluation date(s) for time stream			611 612 613 614 615 CONT SEQ 621 622	combining pooling sequential modeling decision option scanning input iteration PINGENT ANALYSIS INPUT VUENCE values priors
21 CO 21 21 21 21	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex			422 429 VALU 431 432 433	specific other E CRITERIA units base evaluation date(s)			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINGENT ANALYSIS INPUT UNENCE values priors likelihood
21 CO 21 21 21 21 22	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS			422 429 VALU 431 432 433	specific other E CRITERIA units base evaluation date(s) for time stream			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration MINGENT ANALYSIS INPU NUENCE values priors
21 CO 21 21 21 21 22 22 CO	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex			422 429 VALU 431 432 433	specific other E CRITERIA units base evaluation date(s) for time stream ERECT ASSESSMENTS			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINCENT ANALYSIS INPU UNENCE values priors likelihood
21 CO 21 21 21 21 22 22 CO	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS 1 comprehensive			422 429 VALU 431 432 433	specific other E CRITERIA units base evaluation date(s) for time stream			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINCENT ANALYSIS INPU UNENCE values priors likelihood
21 CO 21 21 21 21 22 CO 22 22	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS 1 comprehensive 2 partial			422 429 VALO 431 432 433 INDI 441 442	specific other EC CRITERIA units base evaluation date(s) for time stream RECT ASSESSMENTS conditioned assessment model? Bayesian updating?			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINCENT ANALYSIS INPU UNENCE values priors likelihood
21 CO 21 21 21 21 22 CO 22 22	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS 1 comprehensive			422 429 VALO 431 432 433 INDI 441 442	specific other E CRITERIA units base evaluation date(s) for time stream ERECT ASSESSMENTS conditioned assessment model? Bayesian updating? decomposed assess-			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINGENT ANALYSIS INPUT UNENCE values priors likelihood
21 CO 21 21 21 22 CO 22 22 23 AP	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS 1 comprehensive 2 partial PROXIMATE ANALYSIS?			422 429 VALU 431 432 433 INDI 441 442 443	specific other E CRITERIA units base evaluation date(s) for time stream ERECT ASSESSMENTS conditioned assessment model? Bayesian updating? decomposed assessment model?			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINGENT ANALYSIS INPUT OUENCE values priors likelihood
21 CO 21 21 21 21 22 CO 22 22 23 AP 23 AP	MPLEXITY OF ANALYSIS 1 simple 2 moderately complex 3 very complex MPREHENSIVE/PARTIAL ANALYSIS 1 comprehensive 2 partial			422 429 VALU 431 432 433 INDI 441 442 443 444	specific other E CRITERIA units base evaluation date(s) for time stream ERECT ASSESSMENTS conditioned assessment model? Bayesian updating? decomposed assess-			611 612 613 614 615 CONT SEQ 621 622 623	combining pooling sequential modeling decision option scanning input iteration CINGENT ANALYSIS INPUT OUENCE values priors likelihood

Table B-1: ANALYTIC TAXONOMY - SUMMARY

- User's Options those options that should be considered by the decision maker rather than the analyst.
 - 11 Use decision analysis at all? Should any of the following decision analytic techniques be used or should intuitive decision-making techniques be used?
 - 111 Intuitive Use intuitive decision-making
 devices.
 - 112 Decision analysis Use decision analysis.
 - 119 Other Use another decision-making technique.
 - Dollar amount of analysis What amount of money should be devoted to the analysis effort?
 - 121 Low less than \$15,000.
 - 122 Medium \$15,000 to \$50,000.
 - 123 High greater than \$50,000.
 - Role of decision analysis What role should the analysis serve in the decision-making process?
 - 131 Private/public aid Should the decision analysis serve as a personal aid for the decision maker, or should the analysis serve a more public function?
 - 1311 Private
 - 1312 Public
 - 132 Prescribed/optional Should the decision indicated by the analysis be prescribed, as in an automatic control system, or should the indicated decision be optional?
 - 1321 Prescribed
 - 1322 Optional
 - 133 Contingent/current analysis Should decisions that are anticipated in the future be analyzed in advance, or should they be analyzed only when the decision situation has materialized?

- 1331 Contingent analyzed in advance.
- 1332 Current analyzed when the decision is current.
- Optimization/display only Should a closed optimization model be used [See chapter 18 of Brown, et.al. (1974)], regardless of what display options are used? (Display options are presented in A51 below.)
 - 1341 Optimization
 - 1342 Display only
- 135 Communication Should the analyses serve to communicate the reasons for a choice?
 - 1350 No
 - 1351 Yes
- Organization How should this analysis effort be organized?
 - 141 Analysis source Who should take responsibility for performing the analytic work?
 - 1411 Decision maker
 - 1412 Staff member
 - 1419 Someone else
 - 142 Input source Who should be responsible for providing the model inputs?
 - 1421 Decision maker
 - 1422 Analyst
 - 1423 Expert
 - "Vest pocket" relation to decision maker Should the analyst work in a close, "vest pocket" relation to the decision maker?
 - 1430 No
 - 1431 Yes

- Resources What special resources should be devoted to the analysis?
 - 151 Use a decision-analytic expert? Should a decision analysis expert be used to provide assistance in the analytic effort?

1510 No

1511 Yes

152 Use a computer? - Should a computer be used in the analysis?

1520 No

1521 Yes

- 2 Model Approach Options These options concern the types of decision-analytic modeling techniques to use.
 - 21 Complexity of analysis On the whole, how complex should the analysis effort be?
 - 211 Simple
 - 212 Moderately complex e.g., on a par with a 100-node decision tree.
 - 213 Very complex
 - 22 Comprehensive/Partial analysis Should the analysis encompass only a part of the complete problem or purport to reflect all relevant considerations?
 - 221 Comprehensive reflect all relevant considerations.
 - 222 Partial encompass only a part of the complete problem.
 - 2221 Inference only perform an analysis to draw inferences from the data.
 - 2222 Single value dimension consider only one of the several relevant dimensions of value.
 - 2229 Other perform some other part of the complete analysis.

Approximate analysis - Should the analysis be performed using only approximations of the variables (e.g., values, probabilities, decision options)?

230 No

231 Yes

8

- 3 Input Structure those alternative ways of structuring the inputs to the analysis.
 - 31 Uncertainty How should uncertain events be structured for the analysis?
 - 311 Explicit modeling Is uncertainty explicitly encoded?
 - 3110 No certain equivalents only.
 - 3111 Simple measures of uncertainty e.g., credible intervals.
 - 3112 Complete probability distributions.
 - 312 Time horizon Over what period of time should events be explicitly modeled?

3121 - Short

3122 - Long

- 313 Subsequent acts How should acts subsequent to the initial choice (e.g., an act based upon the outcome of a test marketing effort) be treated in the analysis?
 - 3130 Not explicitly modelled.
 - 3131 Modelled in preposterior format.
 - 3132 Modelled as events.
 - 3139 Other
- 314 Event sequence ordering In what order should the events be structured in the analysis?
 - 3141 Chronological
 - 3149 Non-chronological

315 Detail level - What degree of detail should be included in structuring the inputs?

3151 Low - e.g., binary grouping of variables.

3152 Medium - e.g., grouping by 10 bracket medians.

3153 High

316 Degree of grouping - To what degree should uncertain quantities be grouped?

3161 Low

3162 High

- 32 Value How should valuation be structured in the analysis?
 - 321 Comprehensive Should an attempt be made to use a comprehensive valuation measure, which considers all items of concern to the decision maker?

3210 No

3211 Yes

322 Decomposed - To what degree should the overall value dimension be decomposed into its constituent parts (e.g., decomposing profit into units, price, and profit margin)?

3220 None

3221 Modest

3222 Substantial

323 Partial list - Should only a partial list of value criteria be used in the analysis?

3230 No

3231 Yes

Single index - Should all value dimensions be aggregated into a single index? If so, how should it be done?

3241 Yes

- 32411 Adjustment Treat subsidiary value criteria by making trade-off adjustments to the main criterion.
- 32412 Utility Combine all value criteria into a single utility measure.
- 32419 Other Use another method to arrive at a single valuation index.
- 325 Function If 32412, should multi-attributed value be expressed in the form of a mathematical function? If so, what function?

3250 No

3251 Yes

32511 Linear

32519 Other function

- 33 Special forms Should any special analytical paradigms or techniques be used?
 - 331 Markov Should a sequence of events be treated as a Markov Process?

3310 No

3311 Yes

332 Pareto - Should the concept of Pareto optimality be used in considering decision options?

3320 No

3321 Yes

333 Linear programming - Should linear programming techniques be used to optimize the choice?

3331 Yes

339 Other special forms - Should other special forms of analysis be used?

3390 No

3391 Yes

- Input Specification This category considers how inputs should be specified once they have been structured.
 - 41 Decision options How should the decision option space be explored?
 - 411 Specificity of definition How do the analyzed decision options compare in specificity to the real decision options?
 - 4111 Real option Evaluate the action, e.g., whether to set out for Alaska (with the exact location determined later).
 - 4112 Narrower variant of real option e.g., go to Fairbanks, Alaska.
 - 4113 Broader class e.g., go North.
 - 419 Other
 - 42 Events How should uncertain events be specified in the analysis?
 - 421 Scenarios Incorporate events by specifying scenarios.
 - 422 Specific Use specific events.
 - 429 Other Use another method of specifying events.
 - 43 Value criteria How should the valuation be specified?
 - 431 Units What should the units of value be?
 - 4311 Natural such as dollars, hours, etc.
 - 4319 Other such as utiles.

- 432 Base What kind of reference base should be used for valuation?*
 - 4321 Fixed zero Use an absolute scale.
 - 4322 Floating zero Use a relative or incremental scale (i.e., a scale that relates all values to those of a single, possibly hypothetical, alternative).
- 433 Evaluation date(s) for time stream What evaluation date or dates should be specified in the analysis?
 - 4331 Present value Reduce all values to the present date.
 - 4332 Future value Project all values to a specified future date.
 - 4333 Time flow Present the values as a flow over some specified time period.
 - 4339 Other
- Indirect assessments What indirect assessment techniques should be used to specify inputs?
 - 441 Conditioned assessment the number of conditioning tiers (treating uncertainty explicitly by considering, for each decision option, a probability distribution on the events that impact the valuation).
 - 4410 None
 - 4411 Few
 - 4412 Several
 - 4413 Many
 - 442 Bayesian updating Should Bayes' Theorem be utilized to process information bearing on some model inputs?

^{*}This concept is clarified in Brown, et al. (1974) pp 363-4, 96.

4421 Yes

443 Decomposed assessment model - Should any input be specified by using a decomposed assessment model that breaks the inputs into their constituent parts?

4430 No

4431 Yes

444 Regression - Should any of the inputs be determined by using regression analysis?

4440 No

4441 Yes

449 Other - Should other indirect assessment techniques be used?

4490 No

4491 Yes

- 45 Elicitation technique What elicitation techniques should be used?
 - 451 For discrete probabilities

4511 Odds - Elicit relative likelihoods.

4512 Probabilities - Elicit probabilities.

4519 Other

- 452 For continuous variables
 - 4521 Fixed probability e.g., fractiles or credible intervals.

4522 Fixed interval

4529 Other

453 For values

4531 Reference gamble

4532 Direct rating

4539 Other

454 Use group elicitation - Should group elicitation be used for probabilities or values?

4540 No

4541 Yes

45411 Delphi

45412 Delbecq

45419 Other

5 Output

- 51 Specification What results should be presented?
 - 511 Preferred decision Should the preferred decision be presented?

5110 No

5111 Yes

512 Single value for each option - Should a single value (e.g., an expected value) be presented for each option?

5120 No

5121 Yes

513 Value distributions - Should a distribution of value be presented for each option?

5130 No

5131 Yes

519 Other

52 Display Format - How should the specified output be displayed? 521 Graphic - Should a graphic display of the analytic results be prepared?

5210 No

8

5211 Yes

522 Computer - Should a computer supply the display?

5220 No

5221 Yes

529 Other

- 53 Analytic Devices
 - 531 Use simulation Should the analysis of uncertainty be addressed using Monte Carlo simulation?

5310 No

5311 Yes

53111 Use step-through? - Should the analytical device of step-through simulation be used (See Brown et al., 1974)?

531110 No - Use regular simulation.

531111 Yes

53112 Number of trials - How many simulation trials should be used?

531121 Few

531122 Many

- 539 Other analytic devices What other analytic devices should be used?
 - 5391 Grouping Should grouping devices such as bracket medians be used?

53911 Yes

5392 Pruning - Should pruning devices such as dominance be used?

53920 No

53921 Yes

- 6 Model Management
 - 61 Model Dynamics
 - 611 Combining ~ Should several models be linked together in series?

6110 No

6111 Yes

612 Pooling - Should several simultaneous models be linked together in parallel?

6120 No

6121 Yes

613 Sequential modeling - Should the problem be modeled sequentially?

6130 No

6131 Yes

- 614 Decision option scanning How should the decision option space be scanned?
 - 6141 Complete enumeration by examining all possible decision options.

6142 Sequential

6149 Other

Input iteration - How much should inputs be modified within the same model structure (e.g., sensitivity analysis)?

6151 Values

61511 Low

61512 Medium

61513 High

6152 Probabilities

61521 Low

61522 Medium

61523 High

6153 Both values and probabilities

61531 Low

61532 Medium

61533 High

- 62 Contingent analysis input sequence When should the various inputs be entered into a contingent analysis (Al331 above)?
 - 621 Values When should values be entered?
 - 6211 Early contingently, before the decision is made.
 - 6212 Late when the decision is made.
 - 622 Prior probabilities

6221 Early

6222 Late

623 Likelihoods - Likelihoods for indicators

6231 Early

6232 Late

624 Data

6241 Early

6242 Late

APPENDIX C

8

8

PERFORMANCE MEASURE TAXONOMY

1	Quality of Decision	3 Othe	er Considerations
11	LOGIC OF CHOICE		VITY PRECEDING CHOICE
	111 conceptual completeness 112 effective disaggregation 113 sound predictions 114 good overall logic 115 scope 119 other	311 312 313 314	identification good option generation
12	QUALITY OF INPUT	319	other
	121 good data gathering 122 good management of staff/ expertise		VITY FOLLOWING CHOICE
	123 posing meaningful questio 124 good overall input qualit 129 other		good decision communi- cation good hindsight eval-
2	Time and Cost	323	uation effective implementa- tion
21	ELAPSED TIME	329	
	211 short elapsed modeling 212 fast input assignment 213 fast calculation 214 fast interpretation		ANIZATIONAL AND OTHER -"CHOICE SPECIFIC" ACTS
	215 short overall net elapsed time 219 other	331 332	<pre>improved information improved command, control, and</pre>
22	COSTS	333	communication improved body of applied precepts
		339	other
	inexpensive analysis inexpensive input assignm inexpensive calculation inexpensive overall other	ent	

Table C-1: PERFORMANCE MEASURES — SUMMARY

Quality of decision - the quality of the reasoning or logic and the quality of the inputs.

8

- 11 Logic of choice making choices that are in logical agreement with available information.
 - 111 Conceptual completeness taking all important considerations into account, and avoiding serious errors of approximation.
 - 112 Effective disaggregation dividing the problem into manageable subproblems.
 - 113 Sound predictions making sound inferences from the available data.
 - 114 Good overall logic a summary measure of making choices that are in logical agreement with available information. This includes performing a technically competent analysis.
 - 115 Scope addressing the complete problem (as opposed to addressing only the probability or utility part of the problem).
 - 119 Other logic considerations
- 12 Quality of input obtaining complete, accurate, and timely data and judgments.
 - 121 Good data gathering obtaining complete, accurate, and timely data.
 - 122 Good management of staff/expertise making staff assignments in a manner that makes best use of the available expertise and facilitates accurate and timely data processing.
 - 123 Posing meaningful questions requesting the appropriate data and judgments.
 - 124 Good overall input quality a summary measure of obtaining overall appropriate, complete, accurate, and timely data and judgments.
 - 129 Other input quality considerations

- 2 Time and cost the amount of time and cost devoted to the analysis.
 - 21 Elapsed time the time from the instant that a decision is recognized to the time that a decision is made.
 - 211 Short elapsed modeling time keeping the time spent on structuring the model low.
 - 2111 First pass the first time that the model is used.
 - 2112 Additional passes repeated uses of the model.
 - 212 Fast input assignment quickly attaching values to input parameters (e.g., values and probabilities).
 - 2121 First pass
 - 2122 Additional passes
 - 213 Fast calculation quick performance of the calculations needed to determine a recommended decision.
 - 2131 First pass
 - 2132 Additional passes
 - 214 Fast interpretation ability to quickly interpret the analysis output.
 - 2141 First pass
 - 2142 Additional passes
 - 215 Short overall net elapsed time fast performance of items P211-P214 above.
 - 2151 First pass
 - 2152 Additional passes
 - 2153 Total all passes
 - 219 Other time considerations

- 22 Costs The amount of executive time and anguish and cash expenses involved in making the decision.
 - 221 Low-cost analysis performing an analysis cheaply.

2211 First pass

2212 Additional passes

222 Low-cost input assignment - making all required input assignments cheaply.

2221 First pass

2222 Additional passes

223 Low-cost calculation - performing the required calculations cheaply.

2231 First pass

2232 Additional passes

224 Low-cost overall - a low level of overall
 costs.

2241 First pass

2242 Additional passes

2243 Total - all passes

229 Other costs.

3 Other considerations

- Activity preceding choice processes activities that take place prior to considering a decision.
 - 311 Good environment monitoring monitoring the environment for indications that a problem exists or gathering information to resolve the problem.
 - 312 Good decision identification recognizing when a decision must be made and what decision is needed.
 - 313 Good option generation generating the options for choice.

- 314 Good pre-analysis of anticipated decisions thinking through future decisions in advance.
- 319 Other
- 32 Activity following choice processes.
 - 321 Good decision communication communicating the decision and any supporting arguments (e.g., to facilitate justifying the decision).
 - 322 Good hindsight evaluation of decision facilitating an after-the-fact evaluation of the quality of decision.
 - 323 Effective implementation effectively implementing the choice.
 - 329 Other.
- Organizational and other non-"choice specific" impacts impacts that are not concerned with making a decision choice, but that affect more permanent or organizational properties.
 - 331 Improved information processing improving the recall, correlation, and presentation of relevant data.
 - Improving the command, control, and communication properties of the organization e.g., making managers behave more responsibly.
 - 333 Improving body of applied precepts improving doctrinal guidelines.
 - 339 Other.

APPENDIX D

SITUATIONS FAVORING THE USE OF DECISION ANALYSIS

This appendix presents a more thorough examination of the situations favoring decision analysis than is presented in Section 3.1.2. In particular, this appendix examines how each situation category shown in Tables 2-1 and A-1 impact the decision to use decision analysis. As in other sections of this report, references to situation categories are preceded by the letter "S" and references to performance measures are preceded by the letter "P." These references are explained in detail in Appendices A and C, respectively.

8

Relationship to the Appropriate Amount of Decision Analysis

DECISION SUBSTANCE (S1)

Basic Situation (S11)

Current/Contingent Choice (S111)

Expected Number of Occurrences (S112)

Current choice (S1111) favors the use of decision analysis because it is certain that the analysis could be used. Since decision analysis is a tool to aid in decision making, its value as an aid is reduced in contingent situations (S1112), where it is uncertain that the decision will ever occur.

A decision that is expected to recur (S1123, S1124) supports the use of decision analysis because the cost (P22) per analysis is reduced. The amount of advantage gained by recurrence, of course, depends upon both the expected frequency of occurrence and the similarity of the recurring situations because the similarity will determine the amount of modification needed.

The expected number of occurrences is, in general, closely related to the current/contingent classification, and the two categories should be considered simultaneously. For instance, a current decision that will occur only once should be considered the same as a contingent decision that has a 50% chance of occurring twice. Both of these situations have an expected occurrence rate of one. (The advantages of a recurring situation may be somewhat reduced if the situation is also characterized by hindsight monitoring, (S166) below.)

Operating/Information Act (S113)

Relationship to the Appropriate Amount of Decision Analysis

This classification is unimportant in itself for determining the appropriate amount of analysis, but it is indirectly important because of its impact on other situational dimensions. For example, the maximum option impact (S144) is typically smaller and the determinability of uncertainty (S162) is typically lower for an informational act because the impact of the informational act is felt through a subsequent operating act. This impact relationship obscures the relationship between the situation category and the analytic taxonomy.

Options (S12)

Broad/Narrow (S121)

Clear/Fuzzy (S122)

Complexity of Decision Options (S123)

Radical/Adaptive (S124)

Unimportant

Clearly specified options (S1221) support the use of decision analysis because conceptual completeness is facilitated (P111).

Unimportant

A situation involving a radical decision (S1241) supports the use of decision analysis because decision analysis can improve the logic of choice (P11) in situations that are outside of the decision maker's experience.

Radical decisions also tend to be difficult decisions. This feature is discussed for (S131) below.

Static/Dynamic (S125)

Relationship to the Appropriate Amount of Decision Analysis

Static situations (S1251) favor the use of decision analysis. However, when decision analysis is favored for other reasons in a dynamic situation (S1252), the dynamic feature supports the use of a large amount of analysis. Since decision analysis is not particularly good at providing conceptual completeness (Pll1) in dynamic situations, a good analysis requires a large effort.

Decidability (S13)

Difficulty of Choice (S131)

Difficult choices (S1312) support the use of decision analysis. A basic argument for the use of decision analysis is that it improves the decision quality (P1). In the case of a difficult decision, there is much room for improvement because informal decision processes are not very good at addressing difficult decisions.

Unfamiliarity (S132)

Very unfamiliar decision situations (S1322) support the use of decision analysis mainly because of logic-of-choice considerations (Pl1).

Key Considerations (S133)

If the key consideration is option generation (S1331), decision analysis should not be used. Decision analysis is not good at generating decision options.

If the key consideration is information (S1332), no strong case can be made either in favor of or against decision analysis.

Relationship to the Appropriate
Amount of Decision Analysis

If inference (S1333) is the key consideration, use of decision analysis is slightly favored because decision analysis can help produce sound predictions (P113).

If choice (S1334) is the key consideration, decision analysis is strongly indicated because of its ability to improve choice logic (Pl1).

Stakes (S14)

The stakes involved in the decision is the single most important classification for determining the appropriate amount of decision analysis. Since larger stakes support more decision analysis, only the minimum stakes needed to justify any analysis will be discussed. Alternative definitions of stakes itemized below are in decreasing order of accessibility but in increasing order of relevance. In general, the higher the stakes, the less important is cost (P22) and the stronger the case for decision analysis.

Resources Committed (S141)

R.A. Howard's dictum is that 1% of the amount of resources committed to the decision should be devoted to the decision analysis effort.

Cost Swing (S142)

The cost swing is typically about one half of the resources committed. Accordingly, about 2% of the cost swing should be devoted to the decision analysis effort.

Value Swing (S143)

Relationship to the Appropriate Amount of Decision Analysis

A difference between the best and worst plausible outcomes of about \$1 million is necessary to justify any decision analysis.

Maximum Option Impact (S144)

The maximum option impact is a more relevant but less accessible measure of stakes. As a rough rule of thumb, the maximum option impact, that is, the dollar equivalent difference in outcome attributable to the choice of the best decision over worst decision (among those which might reasonably be taken), should be on the order of \$100,000 (\$1442) in order to justify any decision analysis.

Expected Option Impact (S145)

The expected option impact, that is, the maximum difference in expected value between plausible options under consideration, is an even more relevant and less accessible measure of stakes. An expected option impact of about \$50,000 (S1452) is required to justify the use of decision analysis.

Expected Irrationality Cost (S146)

The most relevant and least accessible measure of stakes is the expected irrationality cost, the room for improvement in the decision (the concept of irrationality cost is explained more fully in Watson and Brown [1975/1]). The expected irrationality cost necessary to support the use of decision analysis is \$20,000.

Relationship to the Appropriate
Amount of Decision Analysis

Difficulty of Net Valuation (S151) Difficult valuation (S1512) supports the use of decision analysis because of the improved logic of choice (Pl1), which decision analysis allows.

The Rest of the Outcome Valuation Dimensions (S151) to (S156) Unimportant.

Outcome Uncertainty (S16)

Number of Uncertainties (S161)

More uncertainties support the use of more decision analysis because of the ability of decision analysis to improve the logic of choice (Pl1) in the face of uncertainty.

Determinability (S162)

The less determinable the distributions, the stronger the case is for using decision analysis, again because of the improved logic of choice (P11) offered by decision analysis.

High/Low Uncertainty (S163)

A high degree of uncertainty supports the use of decision analysis because of the improved logic of choice (Pll) that should result.

Subsequent Acts (S164)

The existence of important subsequent acts supports the use of decision analysis because it can enhance conceptual completeness (Pl1) in that situation.

Type of Evidence (S165)

Unimportant.

Hindsight Monitoring (S166)

A situation that is characterized by a high degree of hindsight monitoring (S1663) promotes improvements in informal

Relationship to the Appropriate
Amount of Decision Analysis

decision making through the development of an improved body of applied precepts (P333). Thus, hindsight monitoring reduces the need to use decision analysis.

DECISION PROCESS (S2)

Reaction Time (S21)

In general, a longer reaction time will support more decision analysis because of the time necessary to perform an analysis (P21). However, a contingent choice analysis (Al331) is supported by a short reaction time combined with a long anticipation time.

Analytic Processes (S22)

Number of Input Sources (S221)

Since conventional, intuitive decision practice has difficulty organizing input from a variety of sources, and decision analysis provides a method for organizing a diverse set of inputs, use of decision analysis can improve the quality of input (P12), especially through the management of staff and expertise (P122); and the logic of choice (Pll), especially through effective disaggregation (P112). Thus, a large number of input sources (S2213, 4, 5) supports the use of a large amount of decision analysis.

Analytic Team (S222)

Decision analysis is favored when a group of people are available to perform the analysis (S22212) mainly because of its capacity for effective disaggregation (P112) and good management of staff and expertise (P122).

Constraints on Analytic Method (S223)

Relationship to the Appropriate
Amount of Decision Analysis

Obviously, if an analytic method is prescribed (\$2231) and the prescribed method is decision analysis, these prescriptions support using decision analysis. On the other hand, if a method of analysis other than decision analysis is prescribed, this argues against using decision analysis. If no constraints are in force (\$2230), this distinction has no effect on the amount of decision analysis to use.

Documentation (S224)

Required documentation favors the use of decision analysis because decision analysis promotes improved communication (P332).

Organizational Processes (S23)

Initiation (S231)

Unimportant.

Responsibility (S232)

Decision analysis is favored in situations in which a tentative decision (S23212) is to be made or where a recommendation is to be made (S23213) because decision analysis can provide good decision communication (P321), effective dissaggregation (P112), and good management of staff and expertise (P122).

Coordination (S233)

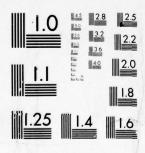
Unimportant.

Justification (S234)

A need to justify the decision (S2341, S2342, or S2343) supports the use of decision analysis because of the need for good decision communication (P321).



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Relationship to the Appropriate
Amount of Decision Analysis

Controversial (S235)

Controversial decisions (S2352) support the use of decision analysis because of the need for good decision communication (P321).

Performance Control (S236)

Tight performance control (S2362) supports the use of decision analysis because of a need for improved command, control, and communication (P332).

Rational-Actor Model (S237)

Since decision analysis assumes a rational-actor model, a situation that is a good approximation of the model (S2372) supports the use of decision analysis.

Risk Attitude (S238)

Decision analysis provides the necessary effective disaggregation (P112) in situations in which risk attitude needs to be considered (S2381, S2382, S2383, or S2389).

Decision Maker Characteristics (S24)

Position in Organization (S241)

A situation with a highlevel decision maker (S24113) supports the use of decision analysis mainly because he is in a position that commands more resources (S25); accordingly, time and cost (P2) are less important considerations.

Personal Characteristics (S242)

Decision analysis is indicated when the decision maker has a familiarity with decision analysis (S24212 or S24213) because in this situation decision analysis will promote effective decision implementation

8

Relationship to the Appropriate
Amount of Decision Analysis

(P323), good management of expertise (P122), and good overall logic (P114) (since familiarity will increase the chances of doing the analysis correctly).

Resources Available (S25)

Computational Facilities (S251)

The use of decision analysis is supported by better computational facilities (S2512) because they enable less expensive calculation (P223).

Staff (S252)

The use of decision analysis is supported by the availability of a strong staff (S2522) because such a staff is more apt to perform an analysis correctly and improve the overall logic of choice (Pl14).

Decision Analysis Expert (S253)

The use of decision analysis is supported by the availability of a decision-analytic expert (S2532) because such an expert is more apt to perform an analysis correctly and improve the overall logic of choice (P114).

Decision Maker Availability (S254)

A high availability of the decision maker (S2542) supports the use of decision analysis because, without it, logic of choice (P11) and quality of input (P12) are seriously impaired.

Assessors Available (S255)

Unimportant.

Dollars Available (S256)

High dollar availability (S2563) favors decision analysis because costs (P22) are then unimportant.

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Other Decision Processes (S29)

Negotiation (S291)

Relationship to the Appropriate Amount of Decision Analysis

Negotiation situations (S2911-2) support the use of decision analysis for reasons of improved choice logic (Pll), especially through conceptual completeness (P11) and effective disaggregation (P112). Decision analysis provides a methodology for simultaneously accounting for elements of the problem that otherwise might be handled individually. (For example, an important amount of work done at Decisions and Designs, Incorporated relates to negotiation problems.)

APPENDIX E

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6

TAXONOMIC MATCHING FOR TYPE OF ANALYSIS

This appendix extends the matching concepts of Section 3.1 to the determination of the form of decision analysis to use. In particular, this appendix characterizes each of the analytic options highlighted in Table 3-2 in terms of performance measure generalizations, and it proposes a tentative matching of analytic options onto situations. The appendix shows how the taxonomies can be used at several illustrative levels of analytic choice, for example, deciding to do conditioned assessment or deciding to use the Delphi technique to aggregate probability judgments. It should be repeated that no attempt has been made at completeness in codifying the state-of-the-art of decision analysis. The matching generalizations presented here are purely illustrative.

Each analytic choice considered will be taken in turn and, in each case, the performance characteristics distinguishing the options will be discussed first, followed by discussion of how various situation characteristics bear on the choice between them.

Throughout this appendix, analytic options references are preceded by the letter "A," situation references are preceded by the letter "S," and performance measure references are preceded by the letter "P." These items are described in detail in Appendix B, Appendix A, and Appendix C.

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E.1 CONTINGENT ANALYSIS (A1331)

Performance Characterization

Contingent choice analysis requires a large amount of first-pass time (P2151)² but provides a fast response on additional passes (P2152). In addition, when reaction time is short (S211), a contingent choice analysis improves both choice logic (P11) and input quality (P12), and improves command, control, and communication (P332) by allowing for incorporation of a senior's values into the analysis.

Situation	
Classificati	on

Current/Contingent Choice (S111)³

Clear/Fuzzy Options (S122)

Expected Number of
Occurrences (S112)
and
Maximum Option
Impact (S114)

Relationship to Contingent Choice Analysis

By definition, contingent analysis requires a contingent choice situation (S1112).

Clear options (S1221) allow a contingent analysis to provide a good pre-analysis of anticipated decision (P314). Clear options also enhance predictability (see below).

These two situational characteristics need to be considered together. In general, contingent choice analysis is indicated in situations where its cost (P22) is justified by either a large number of occurrences (S1124) or by a large maximum option impact (S1444).

¹Analytic options are described in Appendix B.

²Performance measures are described in Appendix C.

³Situation characteristics are described in Appendix A.

As a rule of thumb, a contingent analysis should not be performed unless the expected number of occurrences is at least one (S1122) (e.g., occurring once with certainty or occurring twice with a 50% probability). To determine the number of occurrences it is also necessary to consider the similarity of the occurrences; the more similar the situations are, the more the contingent analysis is justified. (As a rule of thumb, three similar recurrences are roughly equivalent to a single identical recurrence.) It is critical that the actual decision is predictable, in that it involves the same considerations that are modelled for a good pre-analysis (P314).

As a guideline, the expected option impact (considering number of occurrences) should exceed \$10 million (\$1444) to justify the cost (P22) of a contingent analysis. See page 3-3 of Brown et al. (1975) for an illustration of this guideline in a Navy task force commander's decision situation.

Reaction Time (S21)

A contingent analysis is recommended where reaction time is short (S211). In this situation, the analysis can improve the decision quality (P1) by providing a logical framework that considers all available data.

Number of Input Sources (S221)

Contingent choice analyses can improve command, control, and communication (P332) in situations that require several preference (e.g., value) sources to be considered (S2213).

E.2 OPTIMIZATION (A1341)

Performance Characterization

An optimization analysis is characterized by poor decision quality (P1), especially with regard to conceptual completeness (P111), unless the analysis is comprehensive (A221). However, optimization promotes a short interpretation time (P214).

Situation Classification

Reaction Time (S21)

Other

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Relationship to Optimization

If reaction time is short (S211), then the time savings (P214) provided by optimization is more important than the loss in decision quality (P1).

If a comprehensive analysis is performed (A221), then optimization is difficult, leading to lower logic quality (Pll4), and an expensive analysis (P22) if the options are unclear (S1222).

A controversial decision (S2352) may favor display (A1342), but not to the exclusion of optimization (A1341).

E.3 USE A COMPUTER (A1521)

A computerized analysis tends to involve high first-pass calculation cost (P2231) and long first-pass calculation time (P2131), except for a complex analysis (A213). However, a computer analysis generally involves a low level of subsequent-pass calculation cost (P2232) and short subsequent-pass calculation time (P2132).

Situation Classification

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Relationship to Computer Use

Expected Number of Occurrences (S112)

Since a computer offers cheaper and faster performance on subsequent passes, use of a computer is supported by a large expected number of occurrences (S1124).

Maximum Option Impact (S144)

A large maximum option impact (S1444) justifies a large cost (P22) and indicates a complex model (A213). Both of these conditions support a computerized analysis.

Reaction Time (S21)

Since a computer analysis takes a long time to build, days of anticipatory reaction time (S213) are needed. However, since a computerized analysis runs quickly, only minutes of execution reaction time (S211) are needed.

Staff Available (S252)

Some staff support (S2521) is needed to program a computerized analysis.

Other

The costs (P22) and time (P21) of a computerized analysis are not significantly worse than the alternative of a non-computerized analysis if the situation has clear, complex options (S1221, S1236), many determinable uncertainties (S1613, S1623), and many measurable values (S1523, S1532).

A computer is also recommended in conjunction with a complex analysis (A213) and with simulation (A5311).

E.4 COMPLEXITY OF ANALYSIS (A21)

Performance Characterization

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A complex analysis can sometimes (for instance, see below) provide a good choice logic (Pl1) and input quality (Pl2). However, a complex analysis takes a long time (P21) and is expensive (P22). In addition, a complex analysis provides for good decision communication (P321) by explicitly modelling a large number of important factors.

Situation Classification	Relationship to Complexity of Analysis
Difficulty of Choice (S131)	Choice logic (P11) and input quality are important in a difficult choice situation (S1312). If the situation is also one in which a complex analysis contributes to choice logic and input quality, a complex analysis is recommended.
Maximum Option Impact (S144)	A large maximum option impact (S1444) justifies a high cost (P22) to improve choice logic (P11). A complex analysis is supported if it contributes to improved choice logic.
Outcome Valuation Timing (S155)	A complex analysis can improve choice logic (P11) and input quality (P12) if the timing is late (S1552).
Subsequent Acts (S164)	A complex analysis can improve choice logic (P11) and input quality (P12) if subsequent acts have a high impact on the decision outcome (S1643).
Reaction Time (S21)	A complex analysis requires much time (S213 or S214).
Number of Input Sources (S221)	A complex analysis can improve choice logic (P11) and input quality (P12) if there are several input sources (S2213, S2214 or S2215).

Justification (S234)

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A situation that requires justification (S2341, S2342, S2343) requires good decision communication (P321) and thus supports a complex analysis.

Controversial (S235)

Controversial decisions (S2352) require good decision communication (P321) and thus support complex analyses.

Decision Maker's Familiarity with Decision Analysis (S2421) A decision maker who is very familiar with decision analysis (S24213) can improve choice logic (P11) and input quality (P12) by using a complex analysis.

Computational Facilities Availability (S251)

A powerful computer (2512) can reduce the computational cost (P223) of a complex analysis and thus favors its use.

Staff Available (S252)

A strong staff (S2522) is needed to perform a complex analysis in order to minimize the risk of an erroneous analysis and increase the overall choice logic (Pl14).

Decision Maker Availability (S254)

A complex analysis can improve input quality (P12) when a decision maker is unavailable (S2540). The complex analysis can perform the disaggregation that the decision maker might otherwise do informally.

Assessor Availability (S255)

Assessors must be very available (S2552) in order for a complex analysis to provide good overall input quality (P124).

Dollars Available (S256)

Because of its cost (P22), a complex analysis requires a large dollar availability (S2563).

E.5 PARTIAL ANALYSIS - INFERENCE ONLY (A2221)

Performance Generalization

An inference analysis is limited in scope (Pl15) to addressing uncertainty (not values or options). Consequently, an inference analysis costs less (P22) than a full analysis.

Situation Classification	Relation to Inference Only
Key Consideration (S133)	An "inference only" analysis can provide the conceptual completeness (Pll1), effective disaggregation (Pll2), and sound predictions (Pll3) that are necessary when the key consideration is inference (Sl333).
Clear/Fuzzy Options (S122)	An inference model can improve conceptual completeness (Pll1) with fuzzy options (Sl221).
Maximum Option Impact (S144)	An "inference only" analysis is a reduced scope (Pll5) consistent with reduced cost (P22), and thus requires a low threshold of maximum option impact (S1442).
Difficulty of Net Valuation (S151)	Since an "inference only" analysis does not model valuation, conceptual completeness (Pll1) will be poor unless the valuation is easy (S1510).
Uncertainty Determinability (S162)	An inference model can improve choice logic (P11) when uncertainties are indeterminable (S1621).

E.6 APPROXIMATE ANALYSIS (A231)

Performance Generalization

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An approximate analysis can be done quickly (P21) and inexpensively (P22), but it lacks logic of choice (P11) and quality of input (P12).

Situation Classification	Relationship to Approximate Analysis
Difficulty of Choice (S131)	An easy choice (S1311) does not require much improvement in decision quality (P1) and thus supports the use of an approximate analysis.
Maximum Option Impact (S144)	Since an approximate analysis can be cheap (P22), only a low maximum option impact (S1442) is necessary to justify an approximate analysis.
Computational Facilities Available (S251)	An approximate analysis is not complex and is recommended when no computer is available (S2510).
Staff Available (S252)	An approximate analysis is more difficult than a complete one, and thus an approximate analysis requires a strong staff (S2522) in order to control the approximation error and effect good overall choice logic (Pl14).

E.7 SHORT TIME HORIZON (A3121)

Performance Generalization

A short time horizon switches attention from modelling uncertainty to modelling value in that the value model subsumes uncertainties that are beyond the modelled time horizon.

Situation Classification	Relationship to Short Time Horizon
Difficulty of Net Valuation (S151)	Difficult valuation (S1512) supports the use of a short time horizon, since a short horizon focuses attention on valuation.
Valuation Timing (S155)	By definition, a short time horizon should be used in an early timing situation (S1551).
Uncertainty Determinability (S162) and High/Low Uncertainty (P163)	Low (S1631), determinable (1623) uncertainties support the use of a short time horizon, since a short horizon focuses attention away from uncertainty.
Availability of Assessors (S255) and Dollars (S256)	Low assessor availability (S2551) and dollar availability (S2561) will tend to indicate a simple analysis (A211), which can often usefully incorporate a short time horizon.

E.8 ACTS AS EVENTS (A3132)

Performance Characterization

Modelling acts as events reduces cost (P22) and promotes conceptual completeness (P111), but may make posing meaningful questions (P123) difficult compared with preposterior modelling (A3131). However, omitting an explicit modelling of subsequent acts (A3130) reduces cost even more.

Situation Classification

Impact of Subsequent Acts on Outcome (S164)

Other

Relationship to Acts as Events

A high impact of subsequent acts (S1642 or S1643) supports at least some modelling of subsequent acts (but preposterior even more than acts-as-events).

Modelling subsequent acts as events promotes conceptual completeness (Pll1) when it is difficult to model informational events or the subsequent analysis needed for preposterior modelling [see Brown (1975)].

E.9 DECOMPOSED VALUATION (A3222)

Performance Characterization

Sometimes decomposed valuation can improve choice logic (Pll), especially through effective disaggregation (Pll2), and input quality (Pl2). However, decomposed valuation is slow (P21) and expensive (P22). Also, decomposed valuation is limited in scope (Pll5) to valuation.

Situation Classification	Relationship to Decomposed Valuation
Maximum Option Impact (S144)	As a rule of thumb, a decomposed valuation analysis does not require high stakes, but for larger stakes (e.g., S1444) costs (P22) are less important and decomposed valuation is even more strongly indicated.
Difficulty of Net Outcome Valuation (S151)	Decomposed valuation can help provide conceptual completeness (Pll1) and effective disaggregation (Pll2) when outcome valuation is difficult (S1512).
Reaction Time (S21)	Decomposed valuation requires hours of reaction time (S212), but longer reaction time is desirable.
Number of Input Sources (S221)	Decomposed valuation can help effective disaggregation (P112) in a situation with multiple preference sources (S2213 or S2215).
Controversy (S235)	Decomposed valuation can aid good communication (P321) for controversial decisions (S2352).
Computational Facilities (S251)	A computer (S2511 or S2512) is useful, but not necessary, for decomposed valuation.

E.10 SINGLE, ADJUSTED INDEX (A32411)

Performance Characterization

A single, adjusted index promotes good decision communication (P321) because the number is interpretable by a nontechnical user and the number has more information content than a dimensionless utility. If value cannot be reduced to a single index, then insufficient conditions exist for a choice [the situation lacks conceptual completeness (P111)]. An adjusted index is limited in scope (P115) to valuation.

Situation	
Classification	

Adjusted Index

Decision Justification (S234)

An adjusted index can provide the communication (P321) needed to justify a decision (S2341, S2342 or S2343).

Relationship to Single,

Controversy (S235)

An adjusted index can help provide the communication (P321) needed for a controversial decision (S2352).

E.11 LINEAR VALUE FUNCTION (A32511)

Performance Characterization

A linear value function promotes good decision communication (P321) because it is commonly used and readily understood. However, a linear value function lacks conceptual completeness unless certain utility conditions are met. Specifically, value dimensions should be additively independent and linear. A linear value function is limited in scope (P115) to valuation.

Situation Classification	Value Function
Decision Justification (S234)	A linear value function can help provide the communication (P321) needed to justify a decision (S2341, S2342 or S2343).
Staff Available (S252)	Since a linear value function requires no specialized skills, it is indicated when no staff is available (S2520).
Assessor Availability (S255)	Since a linear value function requires no unusually difficult assessments, only a low level of assessor availability (S2551) is needed.

E.12 MARKOV (SPECIAL FORM) (A3311)

Performance Characterization

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A Markov analysis is a compact analysis, which enables inexpensive input assignment (P222) and inexpensive calculation (P223). However, since a Markov analysis is restricted, it may impair conceptual completeness (P111), and, because it requires special skill, a Markov analysis is an expensive analysis (P221). A Markov analysis is also restricted in scope (P115) to uncertainty.

Situation Classification	Relationship to Markov Analysis
Key Consideration (S133)	Markov analysis only addresses probabilities; therefore, it is indicated when inference is the key consideration (S1333).
Difficulty of Net Outcome Valuation (S151)	Since a Markov analysis does not contribute to valuation, it is recommended when valuation is easy (S1510).
Other	A Markov analysis requires a special environmental structure. Specifically, it requires that a time series of stochastically changing discrete variables, and their associated transition availabilities, follow some systematic pattern.

E.13 PARETO (SPECIAL FORM) (A3321)

Performance Characterization

A Pareto analysis can improve conceptual completeness (Pll1) by allowing a simultaneous analysis of the joint utility function of the parties involved. On the other hand, a Pareto analysis may reduce conceptual completeness (Pll1) because it does not accommodate the dynamic social aspects of negotiation.

Situation Classification

Negotiation (S291)

Relationship to Pareto Analysis

Pareto analysis addresses the negotiation problem and thus requires a negotiation situation (S2911 or S2912).

E.14 LINEAR PROGRAMMING (SPECIAL FORM) (A3331)

Performance Characterization

When a computer is available (\$2511 or \$2512), linear programming can provide a low-cost overall analysis (\$P224) and low-cost calculation (\$P223), especially in subsequent passes (\$P2242 and \$P2232), and can enable fast calculation (\$P213) and short elapsed modelling time (\$P211). However, linear programming is slow (\$P21) and expensive (\$P22) without a computer (\$2510). In addition, linear programming will reduce conceptual completeness (\$P111) unless restrictive conditions are met. Specifically, the objective function that is to be maximized or minimized and the constraints must be linear. Also, linear programming can enable low-cost input assignment (\$P222), because it requires little involvement of the decision maker, but it is poor at posing meaningful questions (\$P123) and does not permit disaggregation (\$P112) of value or uncertainties.

Situation Classification	Relationship to Linear Programming
Clear/Fuzzy Options (S122)	Linear programming requires clear options (S1221).
Option Complexity (S123)	Linear programming addresses vector options (S1235 and S1236) and is most useful in large vector (S1236) situations.
Difficulty of Net Outcome Valuation (S151)	Linear programming does not contribute anything to determining valuation and does not permit disaggregation (P112). Thus, the situation must be characterized by easy outcome valuation (S1510).
Amount of Uncertainty (S163)	Linear programming assumes certainty and thus requires a low amount of uncertainty (S1630 or S1631).

Computational Facilities Available (S251) Linear programming requires a computer (S2512) because it is slow (P21) and expensive (P22) without one.

Staff Available (S252)

Linear programming requires a modest staff (S2521), but a strong staff is better (S2522), to ensure an accurate analysis and good overall logic (P114).

Decision Maker Availability (S254)

Linear programming does not require much decision maker availability (S2540) for interaction. This keeps the input assignment costs low (P222).

Other

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Linear programming assumes a structured, linear environment.

E.15 FLOATING ZERO BASE (A4322)

Performance Characterization

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A floating zero base provides a compact analysis. This allows for low-cost input assignment (P222) (because fewer inputs are required), low-cost calculation (P223) and an inexpensive overall analysis (P224). However, a floating zero analysis is less conceptually complete (P111) if utility is non-linear, and is difficult to communicate (P321). In addition, a floating zero analysis requires more assessment skill, which may produce an expensive analysis (P221). Also a floating zero base is limited in scope (P115) to valuation.

Situation Classification	Relationship to Floating Zero Base
Maximum Option Input (S144)	Because floating zero is inexpensive (P22), it requires only a modest maximum option impact (S1442).
Difficulty of Net Outcome Valuation (S151)	Floating zero requires easy valuation (S1510) because it is difficult, and hence expensive (P22), to apply unless valuation is easy.
Decision Justification (S234)	Floating zero is recommended when decision justification is not needed (S2340) because floating zero is not good at providing the communication (P321) needed for justification.
Staff Available (S252)	Floating zero is difficult and thus it requires a strong staff (S2522) to ensure good overall logic (Pll4).

E.16 CONDITIONED ASSESSMENT (A441)

Performance Characterization

Conditioned assessment sometimes promotes logical choice (Pl1), especially through effective disaggregation (Pl12), and quality input (Pl2) (see below). However, conditioned assessment is time consuming (P21) and costly (P22).

Situation Classification	Relationship to Conditioned Assessment
Maximum Option Impact (S144)	The stakes threshold for performing conditioned assessment is low (S1442) but, because of cost of conditioned assessment (P22), the larger the stakes (S144), the stronger the case for using conditioned assessment.
Determinability of Outcome Uncertainty (S162)	Conditioned assessment promotes logical choice (P11), especially through effective disaggregation (P112), and input quality (P12) when outcome uncertainty is indeterminable (S1621).
Reaction Time (S21)	At least hours of reaction time (S212) are needed for conditioned assessment but a longer reaction time is desirable (S213 or S214) because elapsed analysis time (P21) is of less concern with a long reaction time.
Number of Input Sources (S221)	Conditioned assessment can improve input quality (P12), especially through good management of staff/expertise (P122) when there are several substantive sources (S2214 or S2215).
Computational Facility Available (S251)	A computer (S2511, S2512) is useful for conditioned assessment.

Staff Available (S252)

A modest staff (S2521) is necessary but a strong staff is preferred (S2522) in order to ensure good overall logic (Pll4).

E.17 BAYESIAN PROBABILITY UPDATING (A4421)

Performance Characterization

A Bayesian probability updating model promotes effective disaggregation (Pll2) and a low-cost overall analysis (P224). However, a Bayesian updating model may lead to poor input quality (Pl24) because it is often difficult to elicit the necessary prior probabilities and likelihood functions. Also, Bayesian updating is limited in scope (Pll5) to inference.

Situation Classification	Relationship to Bayesian Probability Updating
Key Consideration (S133)	Bayesian probability updating is recommended when the key consideration is inference (S1333) because Bayesian updating addresses inference.
Staff Available (S252)	Bayesian updating is difficult. Therefore, a strong staff (S2522) is needed.
Other	Bayesian updating assumes a perceptual structure that includes prior probabilities and likelihood functions.

E.18 REFERENCE GAMBLE (A4531) (ELICITATION TECHNIQUE FOR VALUES)

Performance Characterization

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The reference gamble elicitation technique can provide conceptual completeness (P111), but the technique makes it difficult to pose meaningful questions (P123). In addition, the reference gamble technique rates low on decision communication (P321), data gathering (P121), and involves high input assignment cost (P222). Also, the reference gamble technique is limited in scope (P115) to valuation.

Situation Classification	Relationship to Reference Gamble Technique
Decision Justification (S234)	Although the reference gamble technique scores low on decision communication (P321), an important performance measure for decision justification, its demonstrably superior conceptual completeness (P111) makes the reference gamble technique good for decision situations requiring justification (S2341, S2342 and S2343).
Staff Available (S252)	Reference gamble elicitation is difficult. Therefore, a strong staff (S2522) is required to ensure good overall logic (P114).
Assessor Availability (S255)	The reference gamble technique requires a high degree of assessor availability (S2552).

E.19 DELPHI (A45411) (GROUP ELICITATION TECHNIQUE FOR PROBABILITIES)

Performance Characterization

The Delphi technique involves pooling the opinions of several probability assessors with limited interaction.

Situation Classification

Number of Input Sources (S221)

Other

Relationship to Delphi Technique

The Delphi technique addresses the problem of several uncertainty sources (S2214).

The Delphi technique limits interaction among assessors and provides the assessors with anonymity. Thus, the Delphi technique removes inhibition and allows better use of staff expertise (P122) where the status of the assessors is heterogeneous. However, if data is heterogeneous, then the Delphi technique reduces input quality (P12) by inhibiting useful interactions. See Fischer (1975) for further information on the Delphi technique.

E.20 SIMULATION (ANALYTIC DEVICE) (A5311)

Performance Characterization

Simulation is expensive when compared with doing nothing, but it provides less costly analysis (P221) and less costly calculation (P223) than an identically structured extensive form analysis. On the other hand, simulation communicates less effectively (P321) than extensive form. In addition, simulation lacks conceptual completeness (P111) due to errors of approximation, and simulation is restricted in scope (P115) to probabilities.

Situation Classification

Relationship to Simulation

Maximum Option Impact (S144)

Since simulation is expensive, it is indicated in situations with large stakes (S1444).

Other

Simulation may be indicated in conjunction with a complex structure (A213) or a conditioned assessment model (A441).

E.21 STEP-THROUGH SIMULATION (A531111) (ANALYTIC DEVICE)

Performance Characterization

Compared with regular simulation (A531110), step-through simulation provides:

- o greater conceptual completeness (P111),
- o lower-cost overall modelling (P224) and input assignment (P222) but greater demands on the decision maker (for the same level of conceptual completeness), and
- o better decision communication (P321) (because step-through requires the decision maker to get involved in the analysis).

Situation	Relationship to Step-
Classification	Through Simulation

Outcome Valuation Timing Step-through simulation is adapted to improving conceptual completeness (P111) for chronological sequences, and chronological

sequences are significant only in situations with late evaluation

dates (S1552).

Determinability of Step-through simulation improves conceptual completeness (Pll1) most when uncertainties are indeterminable (S1621).

Decision Justification
(S234)

Because step-through simulation involves the decision maker in the analysis, it improves the communication (P321) needed for justification (S2341,

S2342, S2343).

Staff Available (S252)

Step-through simulation is a difficult technique, and it requires a strong staff (S2522) to ensure good overall logic (Pl14).

Assessor Availability (S255)

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Step-through simulation can improve conceptual completeness (Pll1) only if the assessor is available (S2552). (Step-through shifts the weight of effort from the analyst to the assessor).

APPENDIX F CONCEPTUAL FRAMEWORK FOR MATCHING

This appendix presents a conceptual framework for the matching task. Specifically, an analogy is drawn between the matching task and the task of formulating a regression problem. This analogy is presented only as a conceptualization of the matching problem and does not imply that the matching task can actually be formulated in a mathematical representation.

Formally, we can visualize the matching task as formulating a regression or other estimating function of the form:

$$\underline{\mathbf{A}} = \mathbf{f} (\underline{\mathbf{p}}) = \overline{\mathbf{q}} (\underline{\mathbf{S}})$$

where

 $\underline{\underline{A}}$ is the appropriate values (choices) of the analytic option vector,

P is the vector of all relevant performance requirements for a given situation,

 \underline{S} is the (much larger) vector of all relevant situation characteristics, and

f and $\overline{\phi}$ are the appropriate mapping functions.

In general, P and S are unavailable because the size of the vector is too great for practicality because the elements in the vector are not accurately measured. Typically, what we have available are P' and S', more manageable sets of estimated performance requirements and situation characteristics, such as those proposed in our performance measure and situation taxonomies.

As in regular regression practice, we are faced with the problem of implicitly or explicitly formulating an estimate of \underline{A} , $\underline{\hat{A}}$, of the form:

$$\hat{A} = \theta \ (P', S')$$

where θ is an <u>estimate</u> of the appropriate mapping function that minimizes the expected error, E(A - A), subject to the cost of measuring and implementing θ , \overline{P} .

The practical counterpart of this approach is that our matching generalizations may incorporate both performance measures and situation characteristics.

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